

Individual life histories: Neither slow nor fast, just diverse

Joanie Van de Walle

23/11/2022

Load packages

```
library(plyr)
library(popbio)
library(vegan)
library(tidyverse)
library(ade4)
library(phytools)
library(MCMCglmm)
library(mulTree)
library(caper)
library(plyr)
library(car)
```

Useful information and functions

```
scl <- 2

scaling <- function(x){
  (x-mean(x))/sd(x)
}

CV_fun <- function(x){
  sd(x)/mean(x)
}

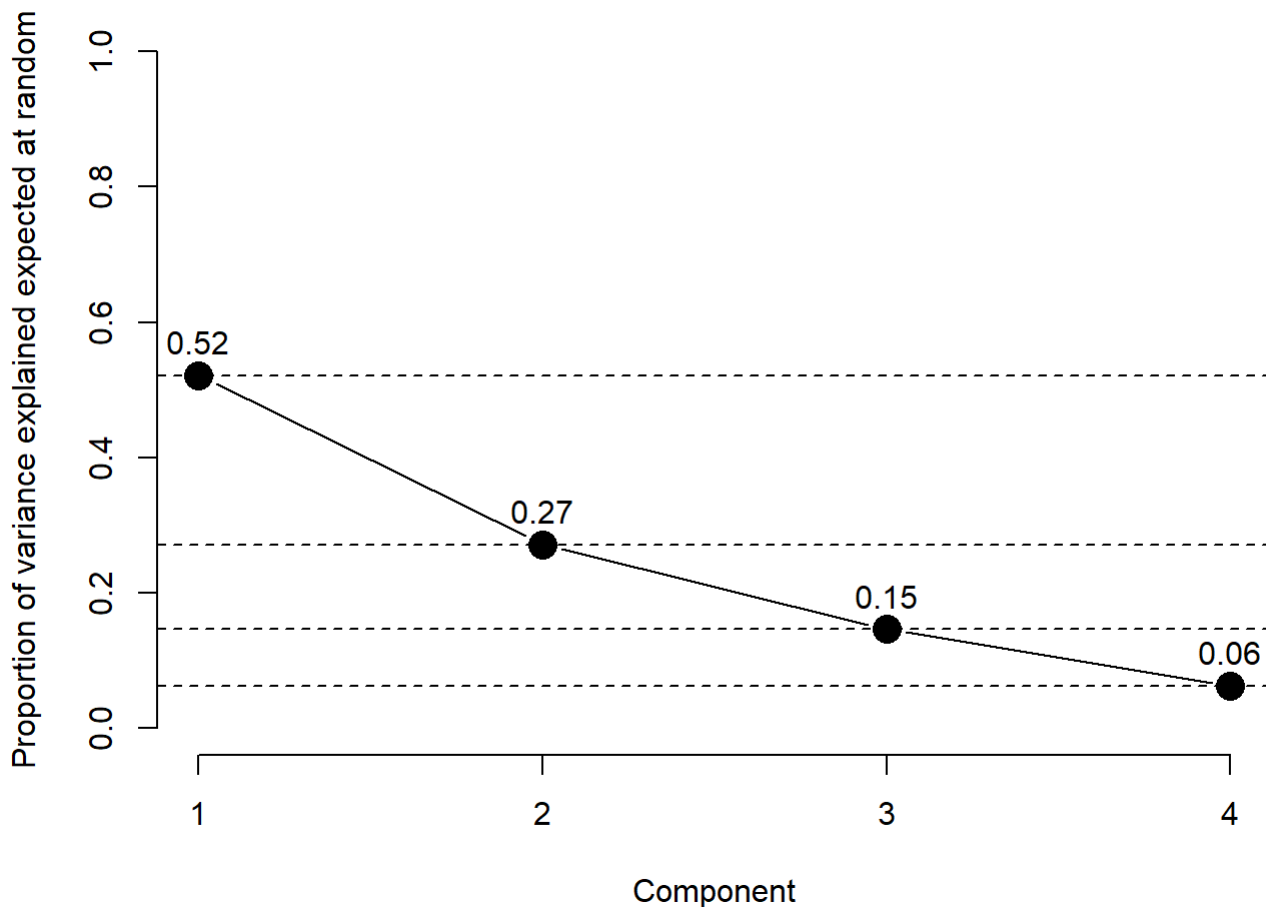
transparency <- 1
colors <- c(rgb(73/255,67/255,166/255,transparency),
            rgb(253/255,231/255,194/255,transparency),
            rgb(3/255,71/255,72/255,transparency),
            rgb(191/255,105/255,0/255,transparency),
            rgb(141/255,36/255,146/255,transparency),
            rgb(173/255,170/255,108/255,transparency),
            rgb(20/255,18/255,67/255,transparency),
            rgb(52/255,143/255,65/255,transparency),
            rgb(224/255,39/255,0/255,transparency),
            rgb(155/255,91/255,100/255,transparency),
            rgb(201/255,228/255,108/255,transparency),
            rgb(131/255,188/255,169/255,transparency),
            rgb(224/255,160/255,0/255,transparency),
            rgb(20/255,100/255,229/255,transparency),
            rgb(255/255,222/255,51/255,transparency),
            rgb(0/255,128/255,167/255,transparency),
            rgb(250/255,18/255,87/255,transparency)
            )

#arrow colors
col_afr <- rgb(18/255,0/255,109/255,0.75)
col_breed <- rgb(194/255,85/255,0/255,0.75)
col_life <- rgb(0/255,117/255,102/255,0.75)
col_fec <- rgb(255/255,214/255,0/255,0.75)
col_points <- rgb(51/255,51/255,51/255,0.5)
```

Broken-stick model

```
# Calculate broken stick number of axes
n = 4 # four variables
bsm = data.frame(j=seq(1:n), p=0)
bsm$p[1] = 1/n
for (i in 2:n) bsm$p[i] = bsm$p[i-1] + (1/(n + 1 - i))
bsm$p = 100*bsm$p/n

par(mfrow=c(1,1), bty="n", mar=c(4,4,2,2))
plot(bsm[c(4:1), 2]/100, pch=16, type="b", xlab="Component", ylab="Proportion of variance explained expected at random", xlim=c(1,4), ylim=c(0,1), cex=2, axes=F)
axis(1, at=seq(1,4,1))
axis(2, at=seq(0,1,0.2))
abline(h=bsm[1,2]/100, lty=2)
text(x=4, y=bsm[1,2]/100+0.05, label=round(bsm[1,2]/100, digits=2))
abline(h=bsm[2,2]/100, lty=2)
text(x=3, y=bsm[2,2]/100+0.05, label=round(bsm[2,2]/100, digits=2))
abline(h=bsm[3,2]/100, lty=2)
text(x=2, y=bsm[3,2]/100+0.05, label=round(bsm[3,2]/100, digits=2))
abline(h=bsm[4,2]/100, lty=2)
text(x=1, y=bsm[4,2]/100+0.05, label=round(bsm[4,2]/100, digits=2))
```



Load dataset

```
#load("../fulldataset.Rdata")
```

```
data <- data_NEE
```

```
head(data) # This dataset contains individual life-histories for the 17 species
```

```
##      species age_first_repro lifetime_adult breed_freq fecundity
## 1 bighornsheep           3           10  1.0000000 0.8000000
## 2 bighornsheep           3            6  1.0000000 1.0000000
## 3 bighornsheep           3           11  0.9090909 0.3636364
## 4 bighornsheep           2           16  1.0000000 0.6923077
## 5 bighornsheep           2           13  0.9230769 0.8461538
## 6 bighornsheep           3            7  0.8571429 0.4285714
##  generation_time
## 1          7.500000
## 2          5.500000
## 3          8.400000
## 4          8.000000
## 5          7.750000
## 6          6.333333
```

Summarise accross species

```
test <- data.frame(
  species_number = 1:17,
  species = sort(unique(data$species)),
  sample_size = ddply(data, .(species), summarise, length(age_first_repro))[,2],
  mean_primiparity = ddply(data, .(species), summarise, mean(age_first_repro))[,2],
  min_primiparity = ddply(data, .(species), summarise, min(age_first_repro))[,2],
  max_primiparity = ddply(data, .(species), summarise, max(age_first_repro))[,2],
  mean_gentime = ddply(data, .(species), summarise, mean(generation_time))[,2],
  min_gentime = ddply(data, .(species), summarise, min(generation_time))[,2],
  max_gentime = ddply(data, .(species), summarise, max(generation_time))[,2],
  mean_adultlife = ddply(data, .(species), summarise, mean(lifetime_adult))[,2],
  min_adultlife = ddply(data, .(species), summarise, min(lifetime_adult))[,2],
  max_adultlife = ddply(data, .(species), summarise, max(lifetime_adult))[,2],
  mean_fecundity = ddply(data, .(species), summarise, mean(fecundity))[,2],
  min_fecundity = ddply(data, .(species), summarise, min(fecundity))[,2],
  max_fecundity = ddply(data, .(species), summarise, max(fecundity))[,2],
  mean_breedingfreq = ddply(data, .(species), summarise, mean(breed_freq))[,2],
  min_breedingfreq = ddply(data, .(species), summarise, min(breed_freq))[,2],
  max_breedingfreq = ddply(data, .(species), summarise, max(breed_freq))[,2])
```

```
table_summary <- test[,c(1,2,3,4,7,10,13,16)] # This dataset contains species averages of life-histories for the 17 species
```

Conduct PCA across species (without phylogeny)

```
# Add 1 to fecundity to avoid having 0 values after log - this is for consistency with the within population level
table_summary$mean_fecundity <- table_summary$mean_fecundity + 1

# Select the life history traits
traits <- c("mean_primiparity", "mean_adultlife", "mean_breedingfreq", "mean_fecundity")

# Perform the PCA
test.pca_all <- rda(log(table_summary[,traits]), scale=T, center=T)

# Summary information
sum_all <- summary(test.pca_all)
sum_all
```

```
##
## Call:
## rda(X = log(table_summary[, traits]), scale = T, center = T)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained    4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1      PC2      PC3      PC4
## Eigenvalue      3.3747 0.4161 0.13631 0.07293
## Proportion Explained 0.8437 0.1040 0.03408 0.01823
## Cumulative Proportion 0.8437 0.9477 0.98177 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 2.828427
##
##
## Species scores
##
##           PC1      PC2      PC3      PC4
## mean_primiparity -1.327 -0.2464 -0.41549 0.06913
## mean_adultlife    -1.375 -0.0083 0.08001 -0.31865
## mean_breedingfreq 1.185 -0.7631 -0.03956 -0.11235
## mean_fecundity     1.301 0.4348 -0.30335 -0.16409
##
##
## Site scores (weighted sums of species scores)
##
##           PC1      PC2      PC3      PC4
## sit1  -0.42818 -0.55696 -0.8911 0.4744
## sit2  -0.08051 -0.75212 0.8018 -0.8565
## sit3  -0.25418 -0.87332 -0.5827 0.8410
## sit4   0.91890 0.52942 -0.1904 -0.4061
## sit5   0.75149 0.14854 0.6302 0.2357
## sit6   0.78226 0.28592 -0.2926 -0.1571
## sit7   1.01402 0.70646 -0.5986 -0.4617
## sit8   0.75835 0.26596 0.2089 -0.2402
## sit9  -1.38028 0.78176 -0.7015 -1.4932
## sit10 -0.34767 0.02015 0.4961 0.1838
## sit11 -0.27735 -1.12036 1.4740 -0.8777
## sit12 -0.92974 1.65903 1.1322 0.9511
## sit13 0.45976 -0.39125 0.3759 1.1612
## sit14 -0.54918 -0.64889 -0.6867 0.3494
## sit15 -0.61357 0.31593 -0.2565 0.6012
```

```
## sit16 -0.26482 -0.33659 -0.3927 -0.1736
## sit17 0.44071 -0.03367 -0.5264 -0.1317
```

```
# Test for proportion of variance explained per axe
prop <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  fec = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop[j,i+1] <- as.numeric(abs(sum_all$species[i,j])/sum(abs(sum_all$species[i,])))
  }
}

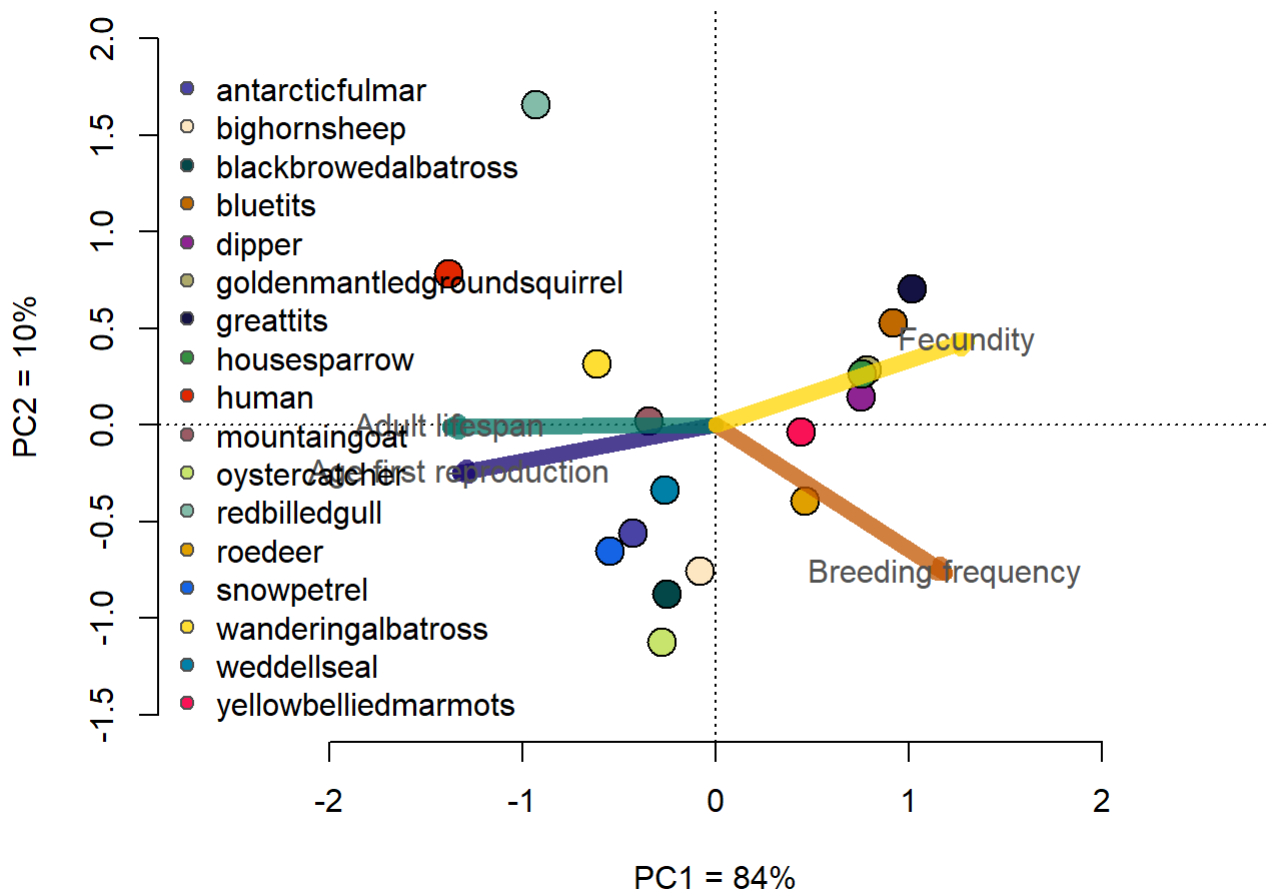
prop
```

```
##   PC      prim      adult      breed      fec
## 1  1 0.64485935 0.771685634 0.56422392 0.59045531
## 2  2 0.11970067 0.004656754 0.36342958 0.19736430
## 3  3 0.20185425 0.044888414 0.01883947 0.13769738
## 4  4 0.03358573 0.178769199 0.05350703 0.07448301
```

```
# Biplot figure
table_summary$species2 <- as.factor(table_summary$species)
species <- with(table_summary, levels(species2))

par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
plot(test.pca_all, scaling = scl, choices=c(1,2), type="n", ylim=c(-1.5,2), xlim=c(-1.5,1.5),
main="", xlab="PC1 = 84%", ylab="PC2 = 10%")
with(table_summary, points(test.pca_all, display = "sites", scaling = scl, pch = 21, bg = color
s[1:17], cex=2))

g <- vegan::scores(test.pca_all, display = "species", choices=1:3)
len <- 1
off_factor <-0
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, c
ol_fec), lwd=8)
text(x=g[1,1]+off_factor, y=g[1,2], labels="Age first reproduction", col = "gray32")
text(x=g[2,1]+off_factor, y=g[2,2], labels="Adult lifespan", col = "gray32")
text(x=g[3,1]+off_factor, y=g[3,2], labels="Breeding frequency", col = "gray32")
text(x=g[4,1]+off_factor, y=g[4,2], labels="Fecundity", col = "gray32")
with(table_summary, legend("bottomleft", legend = species, bty = "n", col = "gray32", pch = 21,
pt.bg = colors))
```



```
# Correlation with generation time
species_score <- vegan::scores(test.pca_all, display = "sites", choices=1)
lm_gen_all <- lm(species_score ~ log10(table_summary$mean_gentime))
summary(lm_gen_all)
```

```
##
## Call:
## lm(formula = species_score ~ log10(table_summary$mean_gentime))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.48902 -0.07074  0.00239  0.13082  0.24124
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.08640    0.08568   12.68 2.03e-09 ***
## log10(table_summary$mean_gentime) -1.53520    0.10332  -14.86 2.22e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1842 on 15 degrees of freedom
## Multiple R-squared:  0.9364, Adjusted R-squared:  0.9321
## F-statistic: 220.8 on 1 and 15 DF, p-value: 2.218e-10
```

pPCA across species - Including phylogeny

```
# Adjust the dataset (create a new one, data_trans) to fit with R functions
data_trans <- table_summary

# Log-transform and scale data prior to PCA (this step is done directly in the rda function for
PCA, but has to be done beforehand in phytools)
trans_list <- c("mean_primiparity",
               "mean_breedingfreq",
               "mean_fecundity",
               "mean_adultlife")

data_trans[,trans_list] <- sapply(data_trans[,trans_list], function(x) log(x))
data_trans[,trans_list] <- sapply(data_trans[,trans_list], function(x) scaling(x))

# Adjust the names of species to match the phylogenetic trees
# Careful as some latin names were changed here compared to the table as synonyms were used in the
# Jetz phylogeny (this is the case for the blue tit and the black-browed albatross)
data_trans$animal <- ifelse(data_trans$species=="bighornsheep","Ovis_canadensis", data_trans$species)
data_trans$animal <- ifelse(data_trans$species=="mountaingoat","Oreamnos_americanus", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="antarcticfulmar","Fulmarus_glacialoides", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="goldenmantledground squirrel","Spermophilus_lateralis", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="dipper","Cinclus_cinclus", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="roedeer","Capreolus_capreolus", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="wanderingalbatross","Diomedea_exulans", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="weddellseal","Leptonychotes_weddellii", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="snowpetrel","Pagodroma_nivea", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="oystercatcher","Haematopusostralegus", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="human","Homo_sapiens", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="housesparrow","Passer_domesticus", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="bluetits","Parus_caeruleus", data_trans$animal)
# Here I had to change this for a synonym.
data_trans$animal <- ifelse(data_trans$species=="blackbrowedalbatross","Thalassarche_melanophrys", data_trans$animal) # here this is a synonym (replace the i by y)
data_trans$animal <- ifelse(data_trans$species=="yellowbelliedmarmots","Marmota_flaviventris", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="greattits","Parus_major", data_trans$animal)
data_trans$animal <- ifelse(data_trans$species=="redbilledgull","Larus_novaehollandiae", data_trans$animal)
```

```
data_trans$species <- data_trans$animal
```

```

# Now Load phylogenetic trees (these were built in the script Phylogenetic Construction_Final)
# axis_trees <- read.tree("../PPCA_100_phylo.tre")
# There are 100 possible trees in this file

# Loop over the 100 trees
res_pca <- list()

for(i in 1:100){
  tree <- axis_trees[[i]]
  tree_name <- paste("tree", i, sep="_")

  data_m <- as.matrix(data_trans[,c(4,6,8,7)])
  row.names(data_m) <- as.factor(data_trans$species)
  Xi <- data_m

  pca<-phyl.pca(tree,Xi, method="lambda")

  ev_1 <-as.numeric(diag(pca$Eval))
  component_importance <- ev_1/sum(pca$Eval)

  prop <- data.frame(
    PC = c(1:4),
    prim = c(1:4),
    adult = c(1:4),
    breed = c(1:4),
    fec = c(1:4)
  )
  for (k in 1:4) {
    for (j in 1:4){
      prop[j,k+1] <- as.numeric(abs(pca$L[k,j])/sum(abs(pca$L[k,])))
    }
  }

  scores_species <- pca$S
  scores_variables <- pca$L

  list_1 <- list(
    tree_name = tree_name,
    pca = pca,
    ev = ev_1,
    component_importance = component_importance,
    variable_correlation = scores_variables,
    scores_species = scores_species,
    prop = prop
  )

  res_pca[[i]] <- list_1
}

ev_all <- data.frame(
  ev1 = rep(0, 100),

```

```

ev2 = rep(0, 100),
ev3 = rep(0, 100),
ev4 = rep(0, 100),
npc = rep(0, 100)
)

prop_all <- data.frame(
  prop1 = rep(0, 100),
  prop2 = rep(0, 100),
  prop3 = rep(0, 100),
  prop4 = rep(0, 100)
)

score_variable_all <- data.frame(
  PC = c(1:4),
  primiparity = rep(0, 4*100),
  adultlife = rep(0, 4*100),
  breed_freq = rep(0, 4*100),
  fecundity = rep(0, 4*100)
)

score_species_all <- data.frame(
  species = rep(data_trans$species, 100),
  PC1 = rep(0, nrow(data_trans)*100),
  PC2 = rep(0, nrow(data_trans)*100),
  PC3 = rep(0, nrow(data_trans)*100),
  PC4 = rep(0, nrow(data_trans)*100))

for(i in 1:100){
  ev_all[i,c(1:4)] <- 100*res_pca[[i]]$ev/sum(res_pca[[i]]$ev)
  prop_all[i,] <- res_pca[[i]]$component_importance
  score_species_all[c((i+16*(i-1)):(i+16*(i-1)+16)), c(2:5)] <- res_pca[[i]]$scores_species[,c
(1:4)]
  score_variable_all[c((i+3*(i-1)):(i+3*(i-1)+3)), c(2:5)] <- res_pca[[i]]$prop[,c(2:5)]
}

# proportion of variance explained by each axe
summary(prop_all[,c(1:4)])

```

##	prop1	prop2	prop3	prop4
##	Min. :0.7646	Min. :0.1040	Min. :0.03161	Min. :0.01823
##	1st Qu.:0.7700	1st Qu.:0.1626	1st Qu.:0.03284	1st Qu.:0.02449
##	Median :0.7741	Median :0.1683	Median :0.03348	Median :0.02472
##	Mean :0.7842	Mean :0.1586	Mean :0.03344	Mean :0.02375
##	3rd Qu.:0.7792	3rd Qu.:0.1715	3rd Qu.:0.03408	3rd Qu.:0.02493
##	Max. :0.8437	Max. :0.1768	Max. :0.03503	Max. :0.02552

Proportion of variance explained by life-history traits

PC1

```
summary(abs(score_variable_all[which(score_variable_all$PC == 1),c(2:5)]))
```

##	primiparity	adulthoodlife	breed_freq	fecundity
##	Min. :0.5523	Min. :0.6227	Min. :0.5642	Min. :0.4726
##	1st Qu.:0.5612	1st Qu.:0.6358	1st Qu.:0.5921	1st Qu.:0.4799
##	Median :0.5648	Median :0.6403	Median :0.5943	Median :0.4837
##	Mean :0.5768	Mean :0.6610	Mean :0.5898	Mean :0.4999
##	3rd Qu.:0.5719	3rd Qu.:0.6519	3rd Qu.:0.5958	3rd Qu.:0.4891
##	Max. :0.6449	Max. :0.7717	Max. :0.5988	Max. :0.5905

PC2

```
summary(abs(score_variable_all[which(score_variable_all$PC == 2),c(2:5)]))
```

##	primiparity	adulthoodlife	breed_freq	fecundity
##	Min. :0.1197	Min. :0.004657	Min. :0.3580	Min. :0.1974
##	1st Qu.:0.1914	1st Qu.:0.103638	1st Qu.:0.3612	1st Qu.:0.2757
##	Median :0.1960	Median :0.112129	Median :0.3624	Median :0.2831
##	Mean :0.1844	Mean :0.095630	Mean :0.3624	Mean :0.2701
##	3rd Qu.:0.1987	3rd Qu.:0.116788	3rd Qu.:0.3634	3rd Qu.:0.2878
##	Max. :0.2049	Max. :0.126970	Max. :0.3671	Max. :0.2962

PC3

```
summary(abs(score_variable_all[which(score_variable_all$PC == 3),c(2:5)]))
```

##	primiparity	adulthoodlife	breed_freq	fecundity
##	Min. :0.2019	Min. :0.04489	Min. :1.853e-05	Min. :0.06906
##	1st Qu.:0.2203	1st Qu.:0.09222	1st Qu.:9.253e-04	1st Qu.:0.08440
##	Median :0.2238	Median :0.09843	Median :2.038e-03	Median :0.09097
##	Mean :0.2214	Mean :0.09159	Mean :4.481e-03	Mean :0.09647
##	3rd Qu.:0.2276	3rd Qu.:0.10460	3rd Qu.:3.238e-03	3rd Qu.:0.09957
##	Max. :0.2331	Max. :0.11583	Max. :1.884e-02	Max. :0.13770

PC4

```
summary(abs(score_variable_all[which(score_variable_all$PC == 4),c(2:5)]))
```

##	primiparity	adulthoodlife	breed_freq	fecundity
##	Min. :0.0008557	Min. :0.1345	Min. :0.03857	Min. :0.07448
##	1st Qu.:0.0089941	1st Qu.:0.1440	1st Qu.:0.04069	1st Qu.:0.13733
##	Median :0.0163090	Median :0.1484	Median :0.04172	Median :0.14355
##	Mean :0.0174141	Mean :0.1518	Mean :0.04335	Mean :0.13350
##	3rd Qu.:0.0266340	3rd Qu.:0.1534	3rd Qu.:0.04325	3rd Qu.:0.14760
##	Max. :0.0335850	Max. :0.1788	Max. :0.05351	Max. :0.15727

Loadings for each axe

```
loadings <- data.frame()
```

```
for (i in 1:100){
  res <- res_pca[[i]]$variable_correlation
  loading <- data.frame(
    PC = 1:4,
    primiparity = as.numeric(res[1,]),
    adulthoodlife = as.numeric(res[2,]),
    breed_freq = as.numeric(res[3,]),
    fecundity = as.numeric(res[4,])
  )
  loadings = rbind(loadings, loading)
}
```

```
summary(abs(loadings[which(loadings$PC == 1), "primiparity"]))
```

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	0.8752	0.8820	0.8841	0.8920	0.8868	0.9386

```
summary(abs(loadings[which(loadings$PC == 2), "primiparity"]))
```

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	0.1742	0.2982	0.3067	0.2867	0.3113	0.3242

```
summary(abs(loadings[which(loadings$PC == 3), "primiparity"]))
```

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	0.2938	0.3480	0.3509	0.3432	0.3552	0.3632

```
summary(abs(loadings[which(loadings$PC == 4), "primiparity"]))
```

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	0.00132	0.01393	0.02557	0.02676	0.04203	0.05330

```
summary(abs(loadings[which(loadings$PC == 1), "adulthoodlife"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.9437 0.9482 0.9500 0.9533 0.9533 0.9726
```

```
summary(abs(loadings[which(loadings$PC == 2), "adulthoodlife"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.00587 0.15187 0.16648 0.14186 0.17412 0.19242
```

```
summary(abs(loadings[which(loadings$PC == 3), "adulthoodlife"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.05658 0.13540 0.14614 0.13442 0.15582 0.17553
```

```
summary(abs(loadings[which(loadings$PC == 4), "adulthoodlife"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.2038 0.2145 0.2198 0.2189 0.2253 0.2291
```

```
summary(abs(loadings[which(loadings$PC == 1), "breed_freq"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.8377 0.8500 0.8521 0.8502 0.8537 0.8565
```

```
summary(abs(loadings[which(loadings$PC == 2), "breed_freq"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.5129 0.5175 0.5198 0.5225 0.5237 0.5396
```

```
summary(abs(loadings[which(loadings$PC == 3), "breed_freq"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 2.655e-05 1.324e-03 2.922e-03 6.579e-03 4.649e-03 2.797e-02
```

```
summary(abs(loadings[which(loadings$PC == 4), "breed_freq"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.05530 0.05829 0.05978 0.06259 0.06202 0.07944
```

```
summary(abs(loadings[which(loadings$PC == 1), "fecundity"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.8113 0.8204 0.8260 0.8399 0.8339 0.9198
```

```
summary(abs(loadings[which(loadings$PC == 2), "fecundity"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.3074 0.4710 0.4837 0.4568 0.4919 0.5074
```

```
summary(abs(loadings[which(loadings$PC == 3), "fecundity"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.1174 0.1442 0.1561 0.1615 0.1705 0.2145
```

```
summary(abs(loadings[which(loadings$PC == 4), "fecundity"]))
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.1160 0.2352 0.2456 0.2263 0.2521 0.2673
```

```
# Correlation with generation time
cor_estimates <- data.frame(
  cor.value = rep(0, 100),
  beta = rep(0, 100),
  SE = rep(0, 100),
  P = rep(0, 100),
  R2 = rep(0, 100)
)

for(i in 1:100){
  tmp <- res_pca[[i]]$scores_species
  score <- as.numeric(tmp[,1])
  gen_time <- table_summary$mean_gentime
  test <- lm(score ~ gen_time)

  cor_estimates$beta[i] <- abs(test$coefficients[2])
  test2 <- summary(test)
  cor_estimates$R2[i] <- test2$r.squared
  cor_estimates$SE[i] = abs(test2$coefficients[2,2])
  cor_estimates$P[i] = test2$coefficients[2,4]

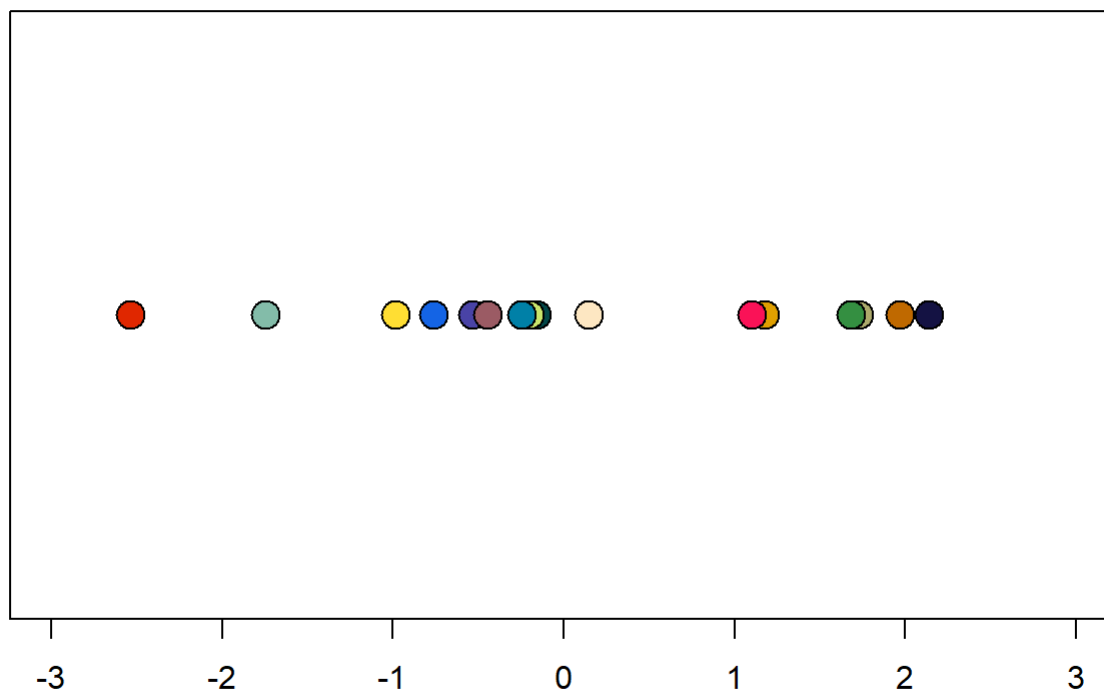
  cor <- cor.test(score, gen_time)
  cor_estimates$cor.value[i] <- abs(cor$estimate)
}

summary(cor_estimates)
```

```
##      cor.value      beta      SE      P
## Min.   :0.8842   Min.   :0.2148   Min.   :0.02924   Min.   :2.340e-06
## 1st Qu.:0.8844   1st Qu.:0.2154   1st Qu.:0.02930   1st Qu.:2.394e-06
## Median :0.8846   Median :0.2156   Median :0.02934   Median :2.419e-06
## Mean    :0.8846   Mean    :0.2160   Mean    :0.02941   Mean    :2.419e-06
## 3rd Qu.:0.8848   3rd Qu.:0.2159   3rd Qu.:0.02938   3rd Qu.:2.452e-06
## Max.    :0.8851   Max.    :0.2188   Max.    :0.02981   Max.    :2.477e-06
##      R2
## Min.   :0.7818
## 1st Qu.:0.7821
## Median :0.7825
## Mean    :0.7825
## 3rd Qu.:0.7828
## Max.    :0.7835
```

```
# Get species' positions on the slow-fast continuum
species_pos <- ddply(score_species_all, .(species), summarize, h = mean(PC1))

# Order species according to their score on PC1
species_pos_new <- species_pos[match(data_trans$species, species_pos$species), ]
stripchart(species_pos_new$h, vertical=F, pch = 21, bg = colors[1:17], cex=2, xlim=c(-3,3))
```



pPCA across species - Including average phylogeny

```
# get the average tree (this takes several minutes - can use the Rdata instead) Note that results may vary according to the average tree computed
# average_tree <- averageTree(axis_trees)
# save(average_tree, file="C:/Users/joani/Documents/PostDoc/Manuscripts/Life history strategies/Intra-specific POL/Data and analyses/Average.phyTree.Rdata")
```

```
#Load("../Average.phyTree.Rdata")
```

```
# Transform data to perform the pPCA - Log transformation and scaling was applied to data_trans above. Transform to matrix format
data_m <- as.matrix(data_trans[,c(4,6,8,7)])
row.names(data_m) <- as.factor(data_trans$species)

# Conduct the pPCA
ppca <- phyl.pca(average_tree, data_m, method="lambda")
sum_ppca <- summary(ppca)
```

```
## Importance of components:
```

```
##           PC1           PC2           PC3           PC4
## Standard deviation    0.1731435 0.08094124 0.03598391 0.03100480
## Proportion of Variance 0.7729192 0.16891236 0.03338400 0.02478447
## Cumulative Proportion 0.7729192 0.94183152 0.97521553 1.00000000
```

```
sum_ppca
```

```
## $sdev
##           PC1           PC2           PC3           PC4
## 0.17314349 0.08094124 0.03598391 0.03100480
##
## $importance
##           PC1           PC2           PC3           PC4
## Standard deviation    0.1731435 0.08094124 0.03598391 0.03100480
## Proportion of Variance 0.7729192 0.16891236 0.03338400 0.02478447
## Cumulative Proportion 0.7729192 0.94183152 0.97521553 1.00000000
##
## attr(,"class")
## [1] "summary.phyl.pca"
```

```
# Test for proportion of variance explained per axe
prop <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  fec = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop[j,i+1] <- as.numeric(abs(ppca$L[i,j])/sum(abs(ppca$L[i,])))
  }
}

prop
```

```
##      PC      prim      adult      breed      fec
## 1  1 0.56357613 0.6400214 0.594226878 0.48299976
## 2  2 0.19694838 0.1121023 0.362378153 0.28348218
## 3  3 0.22526610 0.1010094 0.001929031 0.08888867
## 4  4 0.01420939 0.1468669 0.041465938 0.14462939
```

```
# Correlation with generation time
lm_gen <- lm(as.numeric(ppca$S[, "PC1"])~table_summary$mean_gentime)
summary(lm_gen)
```

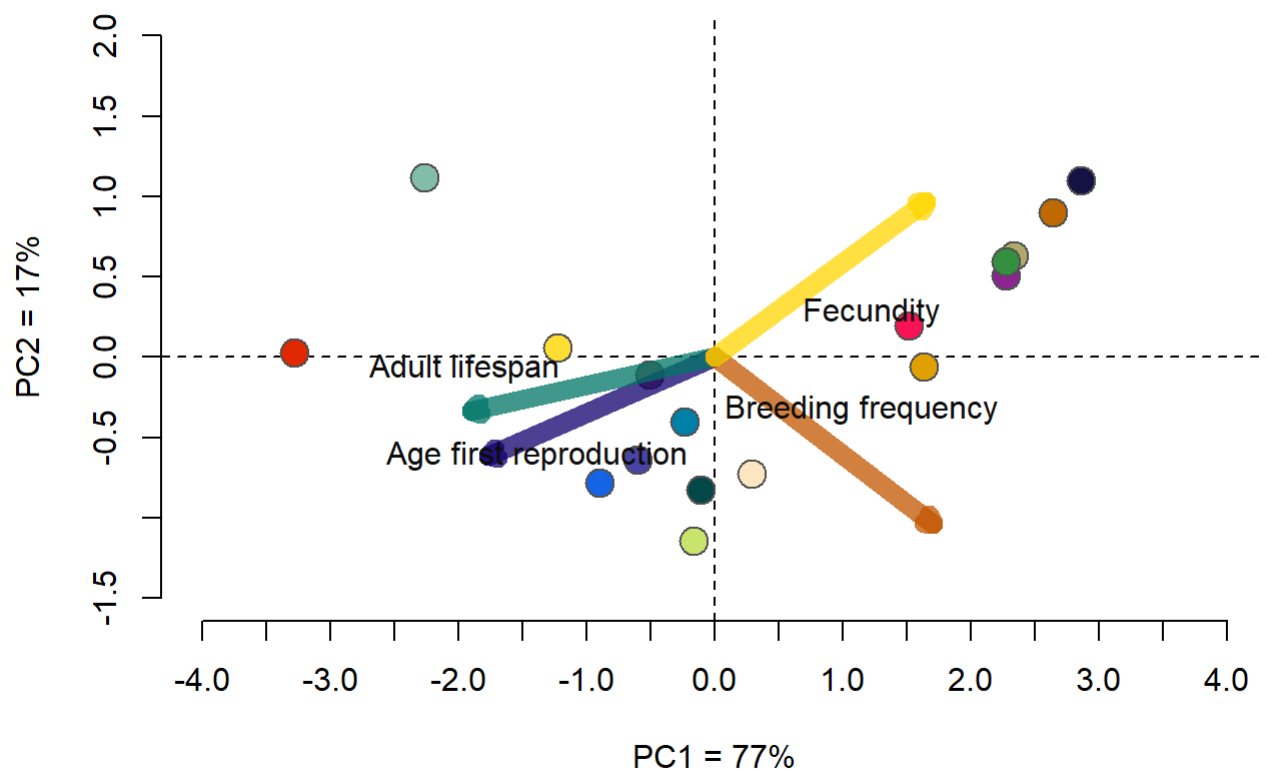
```
##
## Call:
## lm(formula = as.numeric(ppca$S[, "PC1"]) ~ table_summary$mean_gentime)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.2253 -0.4938  0.1362  0.4990  1.1037
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      2.09374     0.31420   6.664 7.56e-06 ***
## table_summary$mean_gentime -0.21555     0.02934  -7.348 2.41e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.8714 on 15 degrees of freedom
## Multiple R-squared:  0.7826, Adjusted R-squared:  0.7681
## F-statistic: 53.99 on 1 and 15 DF, p-value: 2.414e-06
```

```

# Figure
species <- data.frame(ppca$S)
species <- species[,c(1:2)]
species[,2] <- species[,2]*-1
plot(species, type="n", axes=F, ylim=c(-1.5,2), xlim=c(-4,4), main="", xlab="PC1 = 77%", ylab="PC2 = 17%")
axis(1, at=seq(-4,4, 0.5))
axis(2, at=seq(-1.5,2,0.5))
abline(h=0, lty=2)
abline(v=0, lty=2)
points(species, pch = 21, bg = colors[1:17], cex=2, col="gray32")

g <- data.frame(ppca$L)
g <- g[,c(1:2)]
g[,2] <- g[,2]*-1
len <- 2
off_factor <- 0
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=10)
text(x=g[1,1]-0.5, y=g[1,2]-0.3, labels="Age first reproduction", col = "gray5")
text(x=g[2,1]-1, y=g[2,2]+0.1, labels="Adult lifespan", col = "gray5")
text(x=g[3,1]+0.3, y=g[3,2]+0.2, labels="Breeding frequency", col = "gray5")
text(x=g[4,1]+0.4, y=g[4,2]-0.2, labels="Fecundity", col = "gray5")

```



pPCA across species - Including average phylogeny

+ Individual data

```
# Transform individual data for projection on species pPCA
data_trans2 <- data
data_trans2$fecundity <- data_trans2$fecundity +1

# 2 Log-transform and scale life history traits
trans_list2 <- c("age_first_repro",
                "breed_freq",
                "fecundity",
                "lifetime_adult")

data_trans2[,trans_list2] <- sapply(data_trans2[,trans_list2], function(x) log(x)) # don't scale
the individual data at this stage - will do later

# The names of species also have to change to match the phylogenetic trees
# Careful as some latin names were changed here compared to the table as synonyms were used in the
# Jetz phylogeny (this is the case for the blue tit and the black-browed albatross)
data_trans2$animal <- ifelse(data_trans2$species=="bighornsheep","Ovis_canadensis", data_trans2$species)
data_trans2$animal <- ifelse(data_trans2$species=="mountaingoa", "Oreamnos_americanus", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="antarcticfulmar", "Fulmarus_glacialoides", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="goldenmantledground squirrel", "Spermophilus_lateralis", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="dipper", "Cinclus_cinclus", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="roedeer", "Capreolus_capreolus", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="wanderingalbatross", "Diomedea_exulans", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="weddelseal", "Leptonychotes_weddellii", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="snowpetrel", "Pagodroma_nivea", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="oystercatcher", "Haematopus_ostralegus", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="human", "Homo_sapiens", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="housesparrow", "Passer_domesticus", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="bluetits", "Parus_caeruleus", data_trans2$animal) # Here I had to change this for a synonym.
data_trans2$animal <- ifelse(data_trans2$species=="blackbrowedalbatross", "Thalassarche_melanophrys", data_trans2$animal) # here this is a synonym (replace the i by y)
data_trans2$animal <- ifelse(data_trans2$species=="yellowbelliedmarmots", "Marmota_flaviventris", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="greattits", "Parus_major", data_trans2$animal)
data_trans2$animal <- ifelse(data_trans2$species=="redbilledgull", "Larus_novaehollandiae", data_trans2$animal)
```

```

trans2$animal)

data_trans2$species <- data_trans2$animal

# Data set of individual data to be rotated in the pPCA
Xi <- data_trans2[,c("age_first_repro", "breed_freq", "fecundity", "lifetime_adult", "animal")]

# Scale Xi according to species means and SD
mean_afr <- mean(log(table_summary$mean_primiparity))
mean_breed <- mean(log(table_summary$mean_breedingfreq))
mean_fec <- mean(log(table_summary$mean_fecundity + 1))
mean_life <- mean(log(table_summary$mean_adultlife))

sd_afr <- sd(log(table_summary$mean_primiparity))
sd_breed <- sd(log(table_summary$mean_breedingfreq))
sd_fec <- sd(log(table_summary$mean_fecundity + 1))
sd_life <- sd(log(table_summary$mean_adultlife))

Xi$age_first_repro <- (Xi$age_first_repro - mean_afr)/sd_afr
Xi$breed_freq <- (Xi$breed_freq - mean_breed)/sd_breed
Xi$fecundity <- (Xi$fecundity - mean_fec)/sd_fec
Xi$lifetime_adult <- (Xi$lifetime_adult - mean_life)/sd_life

# Get the species in the same order as in the original dataset
Xi$animal <- factor(Xi$animal, levels=data_trans$animal)
Xi <- Xi[order(Xi$animal),]

# transform into matrix format
Xi2 <- as.matrix(Xi[, c(1,4,2,3)])
row.names(Xi2) <- as.factor(Xi$animal)

# Compute the individual scores to be put in the ppca
individual_scores <- scores(ppca,newdata=Xi2)

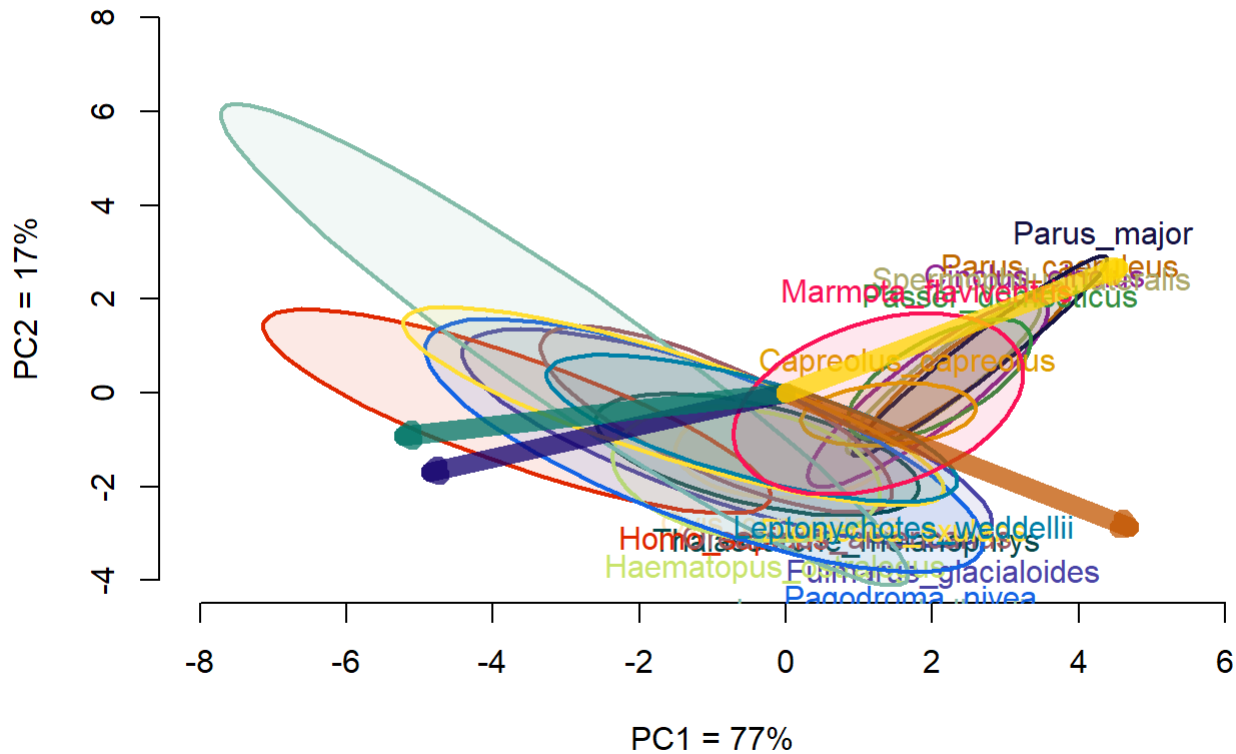
Si <- as.data.frame(individual_scores)
Si$species <- as.factor(Xi$animal)

# Figure
par(bty="n")
plot(Si$PC1, -Si$PC2, type="n", xlim=c(-8, 6), ylim=c(-4,8), main="", xlab="PC1 = 77%", ylab="PC
2 = 17%")
with(Si, dataEllipse(PC1, -PC2, species,
                     xlim=c(-8, 6), center.pch="", pch=16, plot.points=FALSE,
                     ylim=c(-4,8), level=.95, fill=TRUE, fill.alpha=0.1, col=colors))

g3 <- data.frame(ppca$L)
g3 <- g3[,c(1:2)]
g3[,2] <- g3[,2]*-1
len <- 5.5
off_factor <-0

```

```
arrows(0, 0, len * g3[, 1], len * g3[, 2], length = 0.05, col = c(col_afr, col_life, col_breed,
col_fec), lwd=10)
```



```
# Calculate the proportion of variance explained by each axis (G. Blanchet)
# Build result object
varPCAaxis <- matrix(NA,
                     nrow = length(unique(Si$species)),
                     ncol = 4)

colnames(varPCAaxis) <- paste0("PC",1:4)
rownames(varPCAaxis) <- unique(Si$species)

# Calculate the amount of individual variance
for(i in 1:length(unique(Si$species))){
  varPCAaxis[i,] <- apply(Si[which(Si$species == unique(Si$species)[i]),1:4],2,var)
}

# Calculate the proportion of variance associated to each axis
varPCAaxisProp <- varPCAaxis / rowSums(varPCAaxis)

# Final presentation of the results
round(varPCAaxisProp,3)
```

	PC1	PC2	PC3	PC4
## Fulmarus_glacialisoides	0.512	0.209	0.118	0.161
## Ovis_canadensis	0.300	0.143	0.191	0.366
## Thalassarche_melanophrys	0.404	0.140	0.111	0.344
## Parus_caeruleus	0.158	0.214	0.116	0.512
## Cinclus_cinclus	0.144	0.215	0.112	0.530
## Spermophilus_lateralis	0.124	0.189	0.145	0.542
## Parus_major	0.156	0.235	0.120	0.490
## Passer_domesticus	0.172	0.181	0.138	0.509
## Homo_sapiens	0.533	0.211	0.080	0.176
## Oreamnos_americanus	0.441	0.301	0.078	0.180
## Haematopus_ostralegus	0.338	0.208	0.142	0.312
## Larus_novaehollandiae	0.398	0.477	0.031	0.094
## Capreolus_capreolus	0.315	0.098	0.134	0.452
## Pagodroma_nivea	0.476	0.239	0.093	0.193
## Diomedea_exulans	0.623	0.204	0.088	0.085
## Leptonychotes_weddellii	0.546	0.170	0.111	0.172
## Marmota_flaviventris	0.203	0.193	0.129	0.475

Compare CV across species and within populations

```

traits <- c("age_first_repro", "lifetime_adult", "breed_freq", "fecundity")
species <- unique(data$species)

CV <- data.frame(
  species = rep(unique(data$species), each=length(traits)),
  trait = rep(traits, length(unique(data$species))),
  CV = rep(0, length(traits)*length(unique(data$species)))
)

for(j in 1:length(species)){
  for(i in 1:length(traits)){
    CV[which(CV$species == species[j] & CV$trait == traits[i]), "CV"] <- CV_fun(data[which(data
$species == species[j]), traits[i]])
  }
}

# Age at first reproduction
CV_fun(table_summary$mean_primiparity) # across species

```

```
## [1] 1.003706
```

```
mean(CV[which(CV$trait=="age_first_repro"), "CV"]) # Within populations
```

```
## [1] 0.2673112
```

```
# Adult lifespan
CV_fun(table_summary$mean_adultlife) # across species
```

```
## [1] 1.174491
```

```
mean(CV[which(CV$trait=="lifetime_adult"), "CV"]) # Within populations
```

```
## [1] 0.7224116
```

```
# Breeding frequency
CV_fun(table_summary$mean_breedingfreq) # across species
```

```
## [1] 0.1737395
```

```
mean(CV[which(CV$trait=="breed_freq"), "CV"]) # Within populations
```

```
## [1] 0.1440462
```

```
# Fecundity
CV_fun(table_summary$mean_fecundity) # across species
```

```
## [1] 0.6685311
```

```
mean(CV[which(CV$trait=="fecundity"), "CV"]) # Within populations
```

```
## [1] 0.675639
```

Species-specific PCAs

Roe deer

```
data_deer <- data[which(data$species == "roedeer"),]
data_deer$fec_one <- data_deer$fecundity + 1
test.pca_deer1 <- rda(log(data_deer[,c(2,3,4,7)]), scale=TRUE)
summary(test.pca_deer1)
```

```
##
## Call:
## rda(X = log(data_deer[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained   4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      1.616 1.080 0.9177 0.38621
## Proportion Explained 0.404 0.270 0.2294 0.09655
## Cumulative Proportion 0.404 0.674 0.9034 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 3.919476
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro 1.56529 0.04289 -0.92126 -0.7348
## lifetime_adult  0.91794 -0.83077 1.48237 -0.3322
## breed_freq      -1.70605 -0.46592 -0.06994 -0.8414
## fec_one         0.05525 -1.79970 -0.68814 0.3537
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 1298 -0.19297 -0.3942 -0.45697 0.44848
## 2104 -0.19297 -0.3942 -0.45697 0.44848
## 3103 -0.21274 0.5695 -0.02328 -0.08114
## 4103 -0.19297 -0.3942 -0.45697 0.44848
## 5103 0.14289 -0.6480 0.60322 -0.18013
## 6103 -0.19297 -0.3942 -0.45697 0.44848
## 7103 1.12460 -0.5775 -1.16268 -2.06288
## 8103 -0.02291 -0.6245 0.02659 0.19100
## 9103 -0.19297 -0.3942 -0.45697 0.44848
## 10103 0.24382 -0.8536 0.85416 -0.29175
## 11103 0.07082 -0.4793 0.43544 -0.11346
## 1299 -0.19297 -0.3942 -0.45697 0.44848
## 1315 -0.21274 0.5695 -0.02328 -0.08114
## 1415 0.19100 -0.3985 0.90450 -0.44084
## 1515 -0.19297 -0.3942 -0.45697 0.44848
## 1615 -0.19297 -0.3942 -0.45697 0.44848
## 1715 -0.03181 -0.1912 0.22160 -0.04715
```

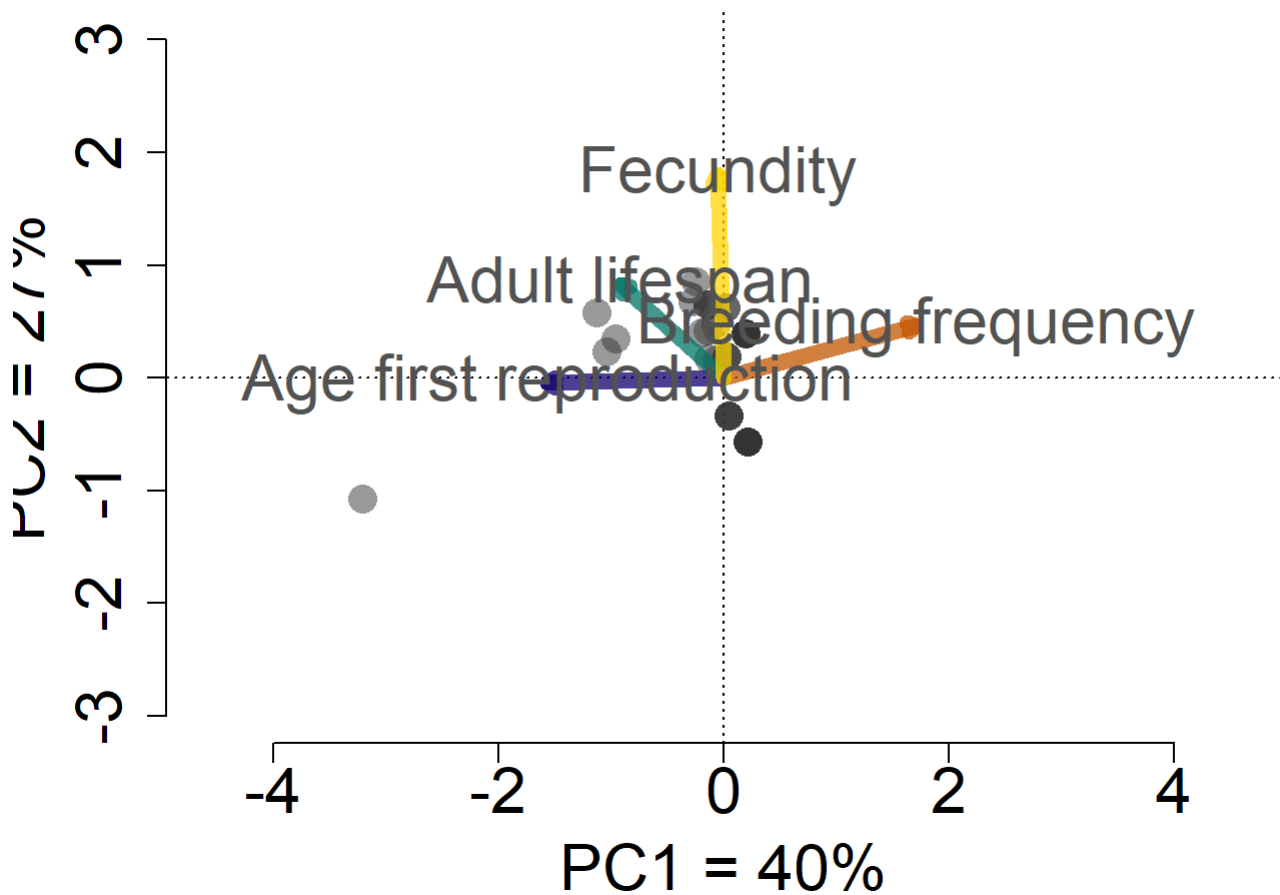
##	1815	0.95455	-0.3472	-1.64624	-1.80540
##	1914	0.13311	-0.1710	0.81786	-0.44225
##	2014	-0.02291	-0.6245	0.02659	0.19100
##	2114	-0.21274	0.5695	-0.02328	-0.08114
##	2214	-0.19297	-0.3942	-0.45697	0.44848
##	2314	-0.21274	0.5695	-0.02328	-0.08114
##	2414	-0.21274	0.5695	-0.02328	-0.08114
##	2514	-0.04269	0.3392	0.46028	-0.33862
##	2613	-0.21274	0.5695	-0.02328	-0.08114
##	2713	-0.19297	-0.3942	-0.45697	0.44848
##	2813	-0.04269	0.3392	0.46028	-0.33862
##	2913	-0.19297	-0.3942	-0.45697	0.44848
##	3013	-0.21274	0.5695	-0.02328	-0.08114
##	3113	-0.04269	0.3392	0.46028	-0.33862
##	3213	-0.21274	0.5695	-0.02328	-0.08114
##	3313	-0.21274	0.5695	-0.02328	-0.08114
##	3413	-0.21274	0.5695	-0.02328	-0.08114
##	3513	-0.21274	0.5695	-0.02328	-0.08114
##	3613	-0.21274	0.5695	-0.02328	-0.08114
##	3713	-0.21274	0.5695	-0.02328	-0.08114
##	3813	-0.21274	0.5695	-0.02328	-0.08114
##	3913	0.14289	-0.6480	0.60322	-0.18013
##	4013	-0.19297	-0.3942	-0.45697	0.44848
##	4113	-0.21274	0.5695	-0.02328	-0.08114
##	4213	-0.21274	0.5695	-0.02328	-0.08114
##	4313	1.03509	-0.2275	1.18151	1.01999
##	4413	-0.04269	0.3392	0.46028	-0.33862
##	4513	-0.21274	0.5695	-0.02328	-0.08114
##	4613	-0.21274	0.5695	-0.02328	-0.08114
##	4713	0.27691	-0.6743	1.06544	-0.47571
##	4813	0.06430	-0.1619	0.57826	-0.28788
##	4913	-0.21274	0.5695	-0.02328	-0.08114
##	5013	-0.03181	-0.1912	0.22160	-0.04715
##	5113	3.20587	1.0695	-0.30163	1.38742
##	5213	-0.19297	-0.3942	-0.45697	0.44848
##	5313	0.14289	-0.6480	0.60322	-0.18013
##	5413	0.13825	-0.4214	0.70517	-0.30463
##	5513	-0.19297	-0.3942	-0.45697	0.44848
##	5613	-0.19297	-0.3942	-0.45697	0.44848
##	5713	-0.03181	-0.1912	0.22160	-0.04715
##	5813	-0.19297	-0.3942	-0.45697	0.44848
##	5913	-0.21274	0.5695	-0.02328	-0.08114
##	6013	-0.21274	0.5695	-0.02328	-0.08114

```

# have the scores reversed to have the adult lifespan to the left
test.pca_deer1_reverse <- test.pca_deer1
test.pca_deer1_reverse$CA$u <- -(test.pca_deer1_reverse$CA$u)
test.pca_deer1_reverse$CA$v <- -(test.pca_deer1_reverse$CA$v)
g <- vegan::scores(test.pca_deer1_reverse, display = "species", scaling=2)

# Figure
par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
plot(test.pca_deer1_reverse, scaling = scl, choices=c(1,2), type="n", ylim=c(-3,3), xlim=c(-3.5,3.5), main="", xlab="PC1 = 40%", ylab="PC2 = 27%", cex.axis=2, cex.lab=2)
with(data_deer, points(test.pca_deer1_reverse, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_deer1_reverse, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)

```



```
# Test for proportion of variance explained per axe
sum_all_deer <- summary(test.pca_deer1)

prop_deer <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  fec = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_deer[j,i+1] <- as.numeric(abs(sum_all_deer$species[i,j])/sum(abs(sum_all_deer$species
[i,])))
  }
}

prop_deer
```

```
##      PC      prim      adult      breed      fec
## 1  1 0.47952887 0.25761377 0.55331403 0.01907323
## 2  2 0.01314097 0.23314890 0.15111073 0.62128278
## 3  3 0.28223168 0.41601621 0.02268376 0.23755606
## 4  4 0.22509848 0.09322112 0.27289148 0.12208792
```

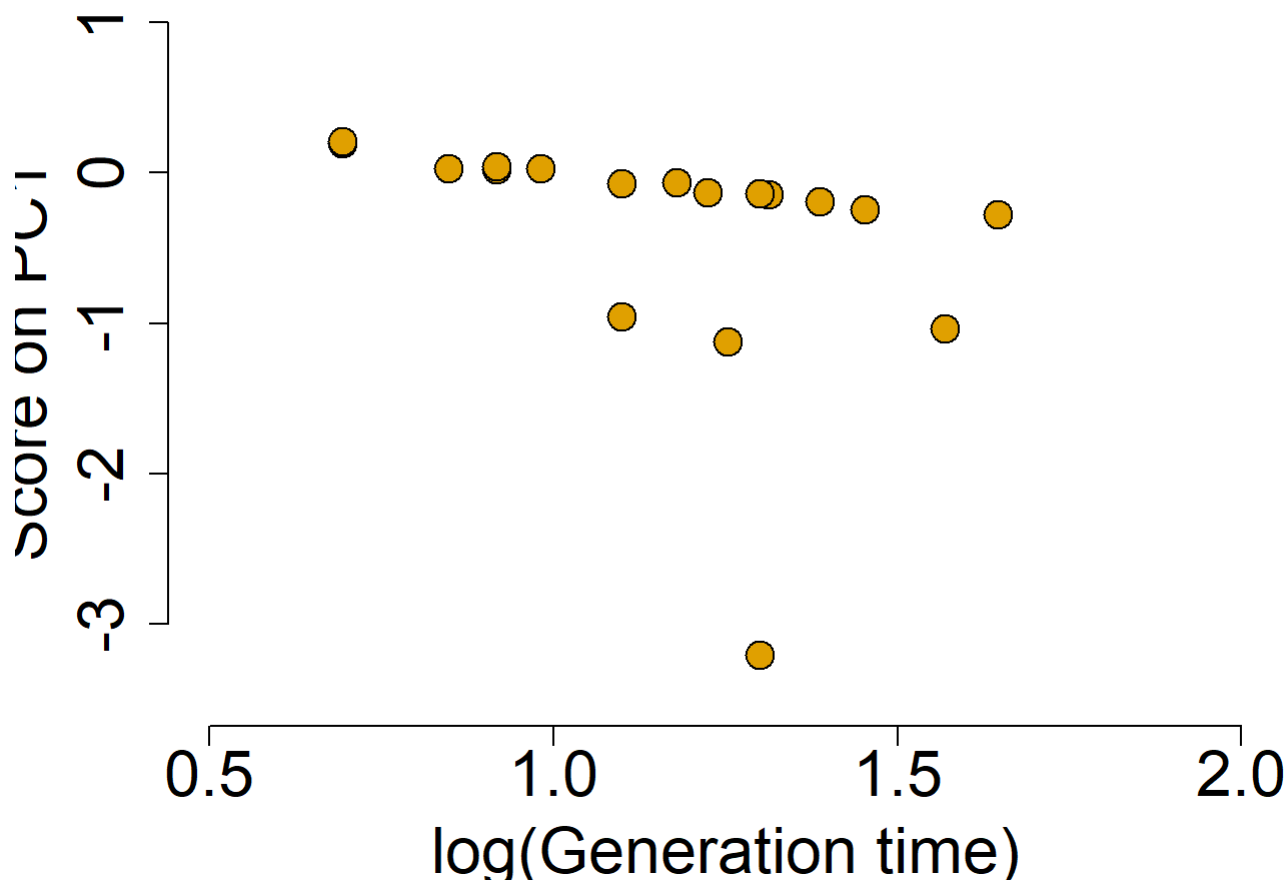
```
# Covariances
test <- as.matrix(scale(log(data_deer[,c(2,3,4,7)])))
cov(test)
```

```
##           age_first_repro lifetime_adult breed_freq   fec_one
## age_first_repro      1.00000000      0.07280684 -0.5227768 0.09982523
## lifetime_adult      0.07280684      1.00000000 -0.2612035 0.10631157
## breed_freq        -0.52277685     -0.26120353  1.0000000 0.12884032
## fec_one           0.09982523      0.10631157  0.1288403 1.00000000
```

```
# Correlation with generation time
individual_score_deer_reverse <- as.numeric(vegan::scores(test.pca_deer1_reverse)$sites[,1])
summary(lm(individual_score_deer_reverse ~ log(data_deer$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_deer_reverse ~ log(data_deer$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.72655 -0.00961  0.01017  0.09115  0.59150
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.9824     0.1817   5.407 1.26e-06 ***
## log(data_deer$generation_time) -1.1250     0.1989  -5.657 4.98e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4132 on 58 degrees of freedom
## Multiple R-squared:  0.3555, Adjusted R-squared:  0.3444
## F-statistic:    32 on 1 and 58 DF,  p-value: 4.982e-07
```

```
par(bty="n")
plot(log(data_deer$generation_time), individual_score_deer_reverse, pch=21, xlab="log(Generation
time)", ylab="Score on PC1", xlim=c(0.5,2), ylim=c(-3.5,1), bg=colors[13], cex=2, cex.axis=2, ce
x.lab=2)
```



Weddell seal

```
data_seal <- data[which(data$species == "weddellseal"),]  
data_seal$fec_one <- data_seal$fecundity + 1  
test.pca_seal1 <- rda(log(data_seal[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_seal1)
```

```
##
## Call:
## rda(X = log(data_seal[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained   4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      2.3971 1.0375 0.5627 0.0027231
## Proportion Explained 0.5993 0.2594 0.1407 0.0006808
## Cumulative Proportion 0.5993 0.8586 0.9993 1.0000000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 4.379877
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro 0.3643 -2.08827 0.5498 0.0004716
## lifetime_adult  1.5456 0.77056 1.3465 0.0036435
## breed_freq      -2.1079 0.11412 0.5772 -0.0797836
## fec_one         -2.1286 0.08909 0.5002 0.0817340
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 1304  0.165885 0.54368 -0.537316 -0.67593
## 2107 -0.269828 0.28069 0.430703 0.48725
## 3106 -0.037224 0.76456 0.435566 0.20346
## 4106 0.318235 -0.63860 0.284304 -0.44804
## 5106 0.421061 -0.77530 -0.302908 -0.65835
## 6106 0.327339 0.05558 0.600207 -0.15527
## 7105 0.654859 -0.14645 -0.384104 -0.30620
## 8105 -0.398290 0.55387 -0.125507 0.24806
## 9105 1.031749 -0.07904 -0.378626 0.63072
## 10105 -0.501537 -0.34229 -0.361975 -0.01671
## 11105 0.241303 -0.45514 -0.052539 -0.59000
## 12101 -0.501537 -0.34229 -0.361975 -0.01671
## 1318 0.319083 -0.05710 0.177890 -0.40855
## 1418 0.051991 -0.13472 1.289042 0.62701
## 1518 0.479940 -0.12695 -0.193964 -0.49844
## 1618 0.037737 0.27472 0.204776 -0.18562
## 1718 0.637464 0.33354 0.030828 -0.22935
```

##	1818	-0.134429	0.08057	1.000370	0.74508
##	1917	0.091270	-0.08390	0.226963	-0.29007
##	2017	0.173887	0.83358	0.557965	0.06561
##	2117	0.352379	0.33734	0.234252	-0.31631
##	2217	-0.444304	-1.10028	0.005911	0.04850
##	2317	-0.325178	0.63809	0.145813	0.39977
##	2417	0.554411	0.31492	0.015198	-0.32076
##	2517	-0.555529	0.37276	-0.709023	-0.07823
##	2616	-0.398290	0.55387	-0.125507	0.24806
##	2716	1.199449	0.46820	-1.511193	1.95636
##	2816	0.384501	-0.08848	0.015384	-0.47845
##	2916	0.411557	0.11547	0.771818	-0.06963
##	3016	1.438815	-1.07739	-0.279173	2.33469
##	3116	0.604029	0.15109	-0.639938	-0.38854
##	3216	0.637179	0.05417	0.389552	-0.17215
##	3316	0.297857	0.46861	0.768719	0.04426
##	3416	0.812536	0.23408	-0.463237	0.04101
##	3516	0.247418	0.29697	0.150498	-0.35285
##	3616	-0.444304	-1.10028	0.005911	0.04850
##	3716	0.252278	0.40039	0.539900	-0.09334
##	3816	0.650191	-0.46028	-0.343213	-0.33594
##	3916	0.253173	0.22307	-0.133905	-0.52217
##	4016	0.112696	-0.36767	0.364691	-0.26566
##	4116	0.319083	-0.05710	0.177890	-0.40855
##	4216	-0.526273	-0.01470	-0.520970	-0.04489
##	4316	0.062473	-0.05287	0.363772	-0.15743
##	4416	0.212654	0.43581	0.700572	0.05477
##	4516	0.273241	-0.41836	0.065981	-0.52372
##	4616	0.149683	0.37420	0.508360	-0.04325
##	4716	-0.369034	0.16642	0.062545	0.28139
##	4816	-0.295922	0.25063	0.333865	0.43311
##	4916	-0.461210	-0.87637	-0.102761	0.02924
##	5016	0.470202	-0.03353	0.167256	-0.37325
##	5116	-0.526273	-0.01470	-0.520970	-0.04489
##	5216	-0.009780	0.40514	0.731236	0.34194
##	5316	0.223386	0.02675	0.559557	-0.13342
##	5416	0.368255	0.39852	0.456034	-0.18956
##	5516	0.397173	0.13380	-0.568318	-0.64705
##	5616	-0.325178	0.63809	0.145813	0.39977
##	5716	-0.021807	0.71091	0.221724	-0.01191
##	5816	-0.344298	-0.16117	0.221541	0.30958
##	5916	-0.257809	-1.30662	0.777479	0.40813
##	6016	-0.345097	-0.98600	0.374069	0.25437
##	6115	0.258790	-0.29151	0.557126	-0.21554
##	6215	0.219876	-0.17137	-0.190268	-0.61441
##	6315	-0.456322	0.48703	-0.340865	0.12763
##	6415	-0.344298	-0.16117	0.221541	0.30958
##	6514	0.862229	0.08178	-1.074922	0.30209
##	6614	0.437967	-0.99921	-0.194236	-0.63909
##	6714	-0.526273	-0.01470	-0.520970	-0.04489
##	6814	0.421908	-0.19379	-0.409322	-0.61886
##	6914	-0.427066	0.09957	-0.152813	0.16097

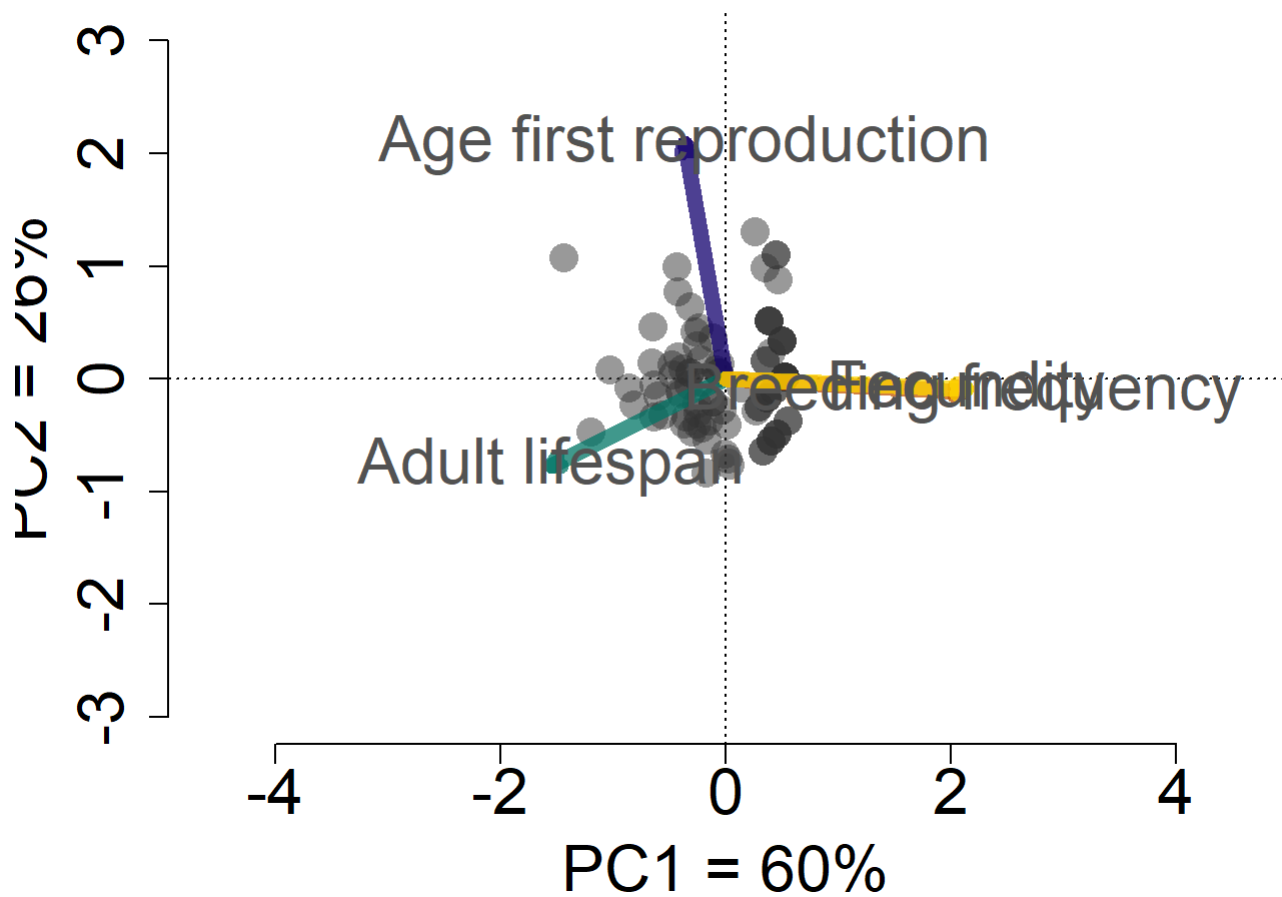
```
## 7013  0.062473 -0.05287  0.363772 -0.15743
## 7113 -0.380904 -0.51179  0.143912  0.21357
## 7213 -0.456322  0.48703 -0.340865  0.12763
## 7313  0.113421  0.20558 -0.107756 -0.47400
## 7413 -0.295922  0.25063  0.333865  0.43311
## 7513 -0.526273 -0.01470 -0.520970 -0.04489
## 7613 -0.456322  0.48703 -0.340865  0.12763
## 7713 -0.380904 -0.51179  0.143912  0.21357
## 7813  0.008482  0.66218  0.016724 -0.21895
## 7913 -0.526273 -0.01470 -0.520970 -0.04489
## 8013 -0.501537 -0.34229 -0.361975 -0.01671
## 8113 -0.501537 -0.34229 -0.361975 -0.01671
## 8212  0.113421  0.20558 -0.107756 -0.47400
## 8313 -0.380904 -0.51179  0.143912  0.21357
## 8413 -0.380904 -0.51179  0.143912  0.21357
## 8512 -0.526273 -0.01470 -0.520970 -0.04489
## 8613 -0.526273 -0.01470 -0.520970 -0.04489
## 8712 -0.402331 -0.22802  0.006183  0.18915
## 8813 -0.526273 -0.01470 -0.520970 -0.04489
## 8913 -0.555529  0.37276 -0.709023 -0.07823
## 9013 -0.369034  0.16642  0.062545  0.28139
## 9113 -0.427066  0.09957 -0.152813  0.16097
## 9212 -0.398290  0.55387 -0.125507  0.24806
## 9312 -0.369034  0.16642  0.062545  0.28139
```

```
# have the scores reversed to have the adult lifespan to the left
```

```
test.pca_seal1_reverse <- test.pca_seal1
test.pca_seal1_reverse$CA$u <- -(test.pca_seal1_reverse$CA$u)
test.pca_seal1_reverse$CA$v <- -(test.pca_seal1_reverse$CA$v)
g <- vegan::scores(test.pca_seal1_reverse, display = "species", scaling=2)
```

```
# Figure
```

```
par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
plot(test.pca_seal1_reverse, scaling = scl, choices=c(1,2), type="n", ylim=c(-3,3), xlim=c(-3.5,3.5), main="", xlab="PC1 = 60%", ylab="PC2 = 26%", cex.axis=2, cex.lab=2)
with(data_seal, points(test.pca_seal1_reverse, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_seal1_reverse, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```
# Test for proportion of variance explained per axe
sum_all_seal <- summary(test.pca_seal1)

prop_seal <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_seal[j,i+1] <- as.numeric(abs(sum_all_seal$species[i,j])/sum(abs(sum_all_seal$species
[i,])))
  }
}

prop_seal
```

```
##      PC      prim      adult      breed      freq
## 1  1 0.1213233197 0.4215642634 0.73216319 0.76030699
## 2  2 0.6954407808 0.2101709692 0.03963798 0.03182304
## 3  3 0.1830788575 0.3672710027 0.20048670 0.17867584
## 4  4 0.0001570421 0.0009937648 0.02771212 0.02919413
```

```
# Covariances
```

```
test <- as.matrix(scale(log(data_seal[,c(2,3,4,7)])))
cov(test)
```

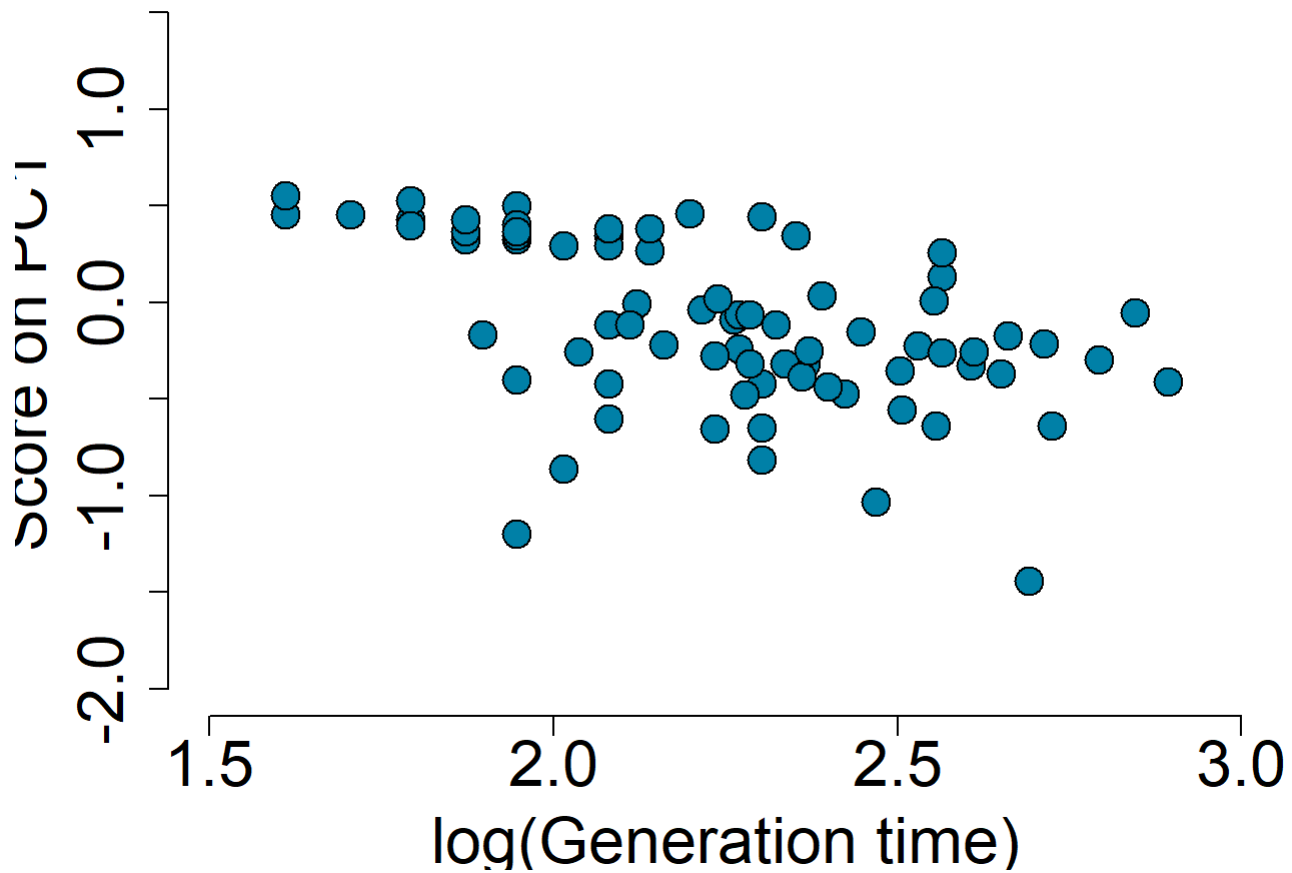
```
##      age_first_repro lifetime_adult breed_freq fec_one
## age_first_repro      1.00000000      -0.06376304 -0.1436586 -0.1431420
## lifetime_adult      -0.06376304      1.00000000 -0.4989979 -0.5311788
## breed_freq          -0.14365862      -0.49899789  1.0000000  0.9965525
## fec_one             -0.14314203      -0.53117878  0.9965525  1.0000000
```

```
# Correlation with generation time
```

```
individual_score_seal_reverse <- as.numeric(vegan::scores(test.pca_seal1_reverse)$sites[,1])
summary(lm(individual_score_seal_reverse ~ log(data_seal$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_seal_reverse ~ log(data_seal$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.40388 -0.17364  0.06127  0.21145  0.60950
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.9526     0.2642   7.391 6.85e-11 ***
## log(data_seal$generation_time) -0.8984     0.1203  -7.466 4.80e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3616 on 91 degrees of freedom
## Multiple R-squared:  0.3799, Adjusted R-squared:  0.3731
## F-statistic: 55.75 on 1 and 91 DF, p-value: 4.804e-11
```

```
par(bty="n")
plot(log(data_seal$generation_time), individual_score_seal_reverse, pch=21, xlab="log(Generation
time)", ylab="Score on PC1", xlim=c(1.5,3), ylim=c(-2,1.5), bg=colors[16], cex=2, cex.axis=2, ce
x.lab=2)
```



Golden-mantled ground squirrel

```
data_squirrels <- data[which(data$species == "goldenmantledgroundsquirrel"),]  
data_squirrels$fec_one <- data_squirrels$fecundity + 1  
test.pca_squirrels1 <- rda(log(data_squirrels[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_squirrels1)
```

```
##
## Call:
## rda(X = log(data_squirrels[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           3           1
## Unconstrained   3           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1      PC2      PC3
## Eigenvalue      1.347 1.0180 0.6349
## Proportion Explained 0.449 0.3393 0.2116
## Cumulative Proportion 0.449 0.7884 1.0000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 3.707793
##
##
## Species scores
##
##           PC1      PC2      PC3
## age_first_repro 0.2574 -2.0863 0.4043
## lifetime_adult  1.7239 0.5204 1.1576
## breed_freq      0.0000 0.0000 0.0000
## fec_one         1.7706 -0.2033 -1.1858
##
##
## Site scores (weighted sums of species scores)
##
##           PC1      PC2      PC3
## 1293    0.69904 -0.4609 0.35430
## 2101    0.23207 0.4732 0.62979
## 3101   -0.44682 0.2304 -0.19093
## 4101   -0.19717 0.1925 -0.54566
## 5101    0.11416 -0.7180 -0.60001
## 6101    0.04890 -0.7081 -0.50728
## 7101   -0.22459 -0.6665 -0.11866
## 8101   -0.90210 -0.5635 0.84405
## 9101    0.04890 -0.7081 -0.50728
## 10101   0.40853 -0.5444 0.10816
## 11101  -0.14008 0.4021 0.49966
## 1294    0.55978 -1.3029 0.94023
## 1312   -0.11554 -0.6831 -0.27362
## 1412   -0.36519 -0.6451 0.08112
## 1512   -0.56335 -0.6150 0.36270
## 1612    0.40853 -0.5444 0.10816
## 1712   -0.02643 -0.6966 -0.40023
```

## 1812	0.32689	-0.5320	0.22417
## 1911	0.46177	0.4383	0.30340
## 2011	0.28778	0.3371	-0.10830
## 2111	-0.19717	0.1925	-0.54566
## 2211	-0.30623	0.2091	-0.39070
## 2311	-0.98374	0.3120	0.57200
## 2411	0.28778	0.3371	-0.10830
## 2511	0.43023	0.4431	0.34822
## 2611	0.32690	0.3311	-0.16388
## 2711	0.59467	0.4181	0.11456
## 2811	0.32690	0.3311	-0.16388
## 2911	-0.11554	-0.6831	-0.27362
## 3011	0.36312	0.3256	-0.21535
## 3111	0.46177	0.4383	0.30340
## 3211	-0.30623	0.2091	-0.39070
## 3311	-0.02643	-0.6966	-0.40023
## 3411	-0.19717	0.1925	-0.54566
## 3511	0.19868	0.3506	0.01831
## 3611	-0.19717	0.1925	-0.54566
## 3711	-0.02643	-0.6966	-0.40023
## 3811	-0.19717	0.1925	-0.54566
## 3911	-0.19717	0.1925	-0.54566
## 4011	-0.98374	0.3120	0.57200
## 4111	0.32118	0.4597	0.50318
## 4211	0.65993	0.4082	0.02183
## 4311	-0.90210	-0.5635	0.84405
## 4411	0.96803	-0.3409	0.80224
## 4511	0.61837	0.5051	0.54841
## 4611	-0.30623	0.2091	-0.39070
## 4711	0.24526	0.3435	-0.04788
## 4811	0.28031	-0.5249	0.29036
## 4911	-0.98374	0.3120	0.57200
## 5011	0.43023	0.4431	0.34822
## 5111	0.08962	0.3672	0.17327
## 5211	0.03253	0.1576	-0.87205
## 5311	-0.98374	0.3120	0.57200
## 5411	0.65246	-0.4538	0.42049
## 5511	-0.30623	0.2091	-0.39070
## 5611	-0.14008	0.4021	0.49966
## 5711	-0.98374	0.3120	0.57200
## 5811	0.36312	0.3256	-0.21535
## 5911	-0.19717	0.1925	-0.54566
## 6011	0.46177	0.4383	0.30340
## 6111	-0.10807	0.1789	-0.67228
## 6211	0.14718	0.3584	0.09148
## 6311	-0.03273	0.1675	-0.77932
## 6411	-0.19717	0.1925	-0.54566

```

# have the scores reversed to have the adult lifespan to the left
test.pca_squirrels1_reverse <- test.pca_squirrels1
test.pca_squirrels1_reverse$CA$u <- -(test.pca_squirrels1_reverse$CA$u)
test.pca_squirrels1_reverse$CA$v <- -(test.pca_squirrels1_reverse$CA$v)
g <- vegan::scores(test.pca_squirrels1_reverse, display = "species", scaling=2)

# Figure
par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
plot(test.pca_squirrels1_reverse, scaling = scl, choices=c(1,2), type="n", ylim=c(-3,3), xlim=
c(-3.5,3.5), main="", xlab="PC1 = 45%", ylab="PC2 = 34%", cex.axis=2, cex.lab=2)
with(data_squirrels, points(test.pca_squirrels1_reverse, display = "sites", col = col_points, sc
aling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_squirrels1_reverse, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, c
ol_fec), lwd=8)

```

```

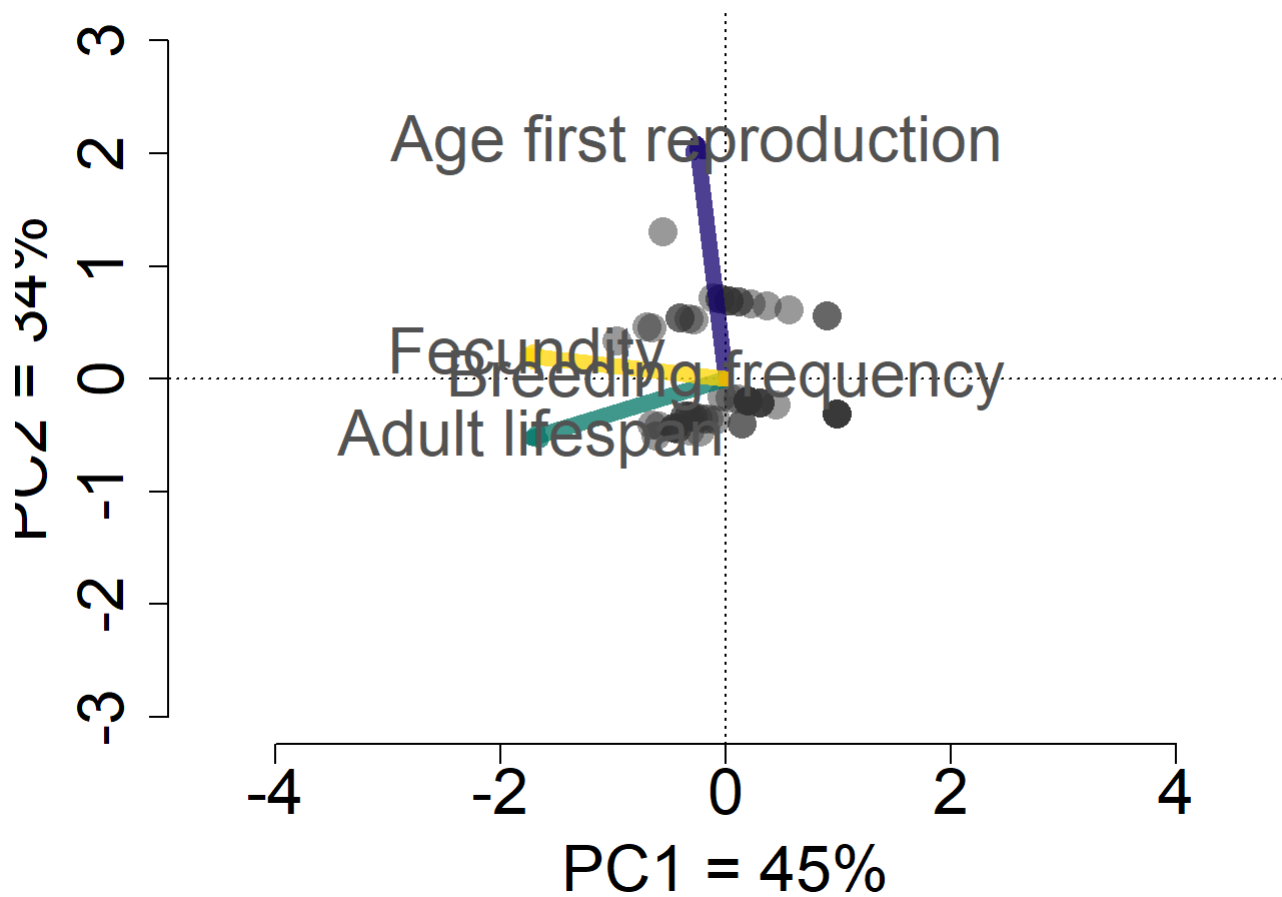
## Warning in arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col =
## c(col_afr, : zero-length arrow is of indeterminate angle and so skipped

```

```

text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)

```



```
# Test for proportion of variance explained per axe
sum_all_squirrels <- summary(test.pca_squirrels1)

prop_squirrels <- data.frame(
  PC = c(1:3),
  prim = c(1:3),
  adult = c(1:3),
  breed = c(1:3),
  freq = c(1:3)
)

for (i in 1:4) {
  for (j in 1:3){
    prop_squirrels[j,i+1] <- as.numeric(abs(sum_all_squirrels$species[i,j])/sum(abs(sum_all_squirrels$species[i,])))
  }
}

prop_squirrels
```

```
##   PC      prim    adult breed      freq
## 1  1 0.09367063 0.5067541   NaN 0.56037105
## 2  2 0.75920747 0.1529744   NaN 0.06435582
## 3  3 0.14712191 0.3402715   NaN 0.37527313
```

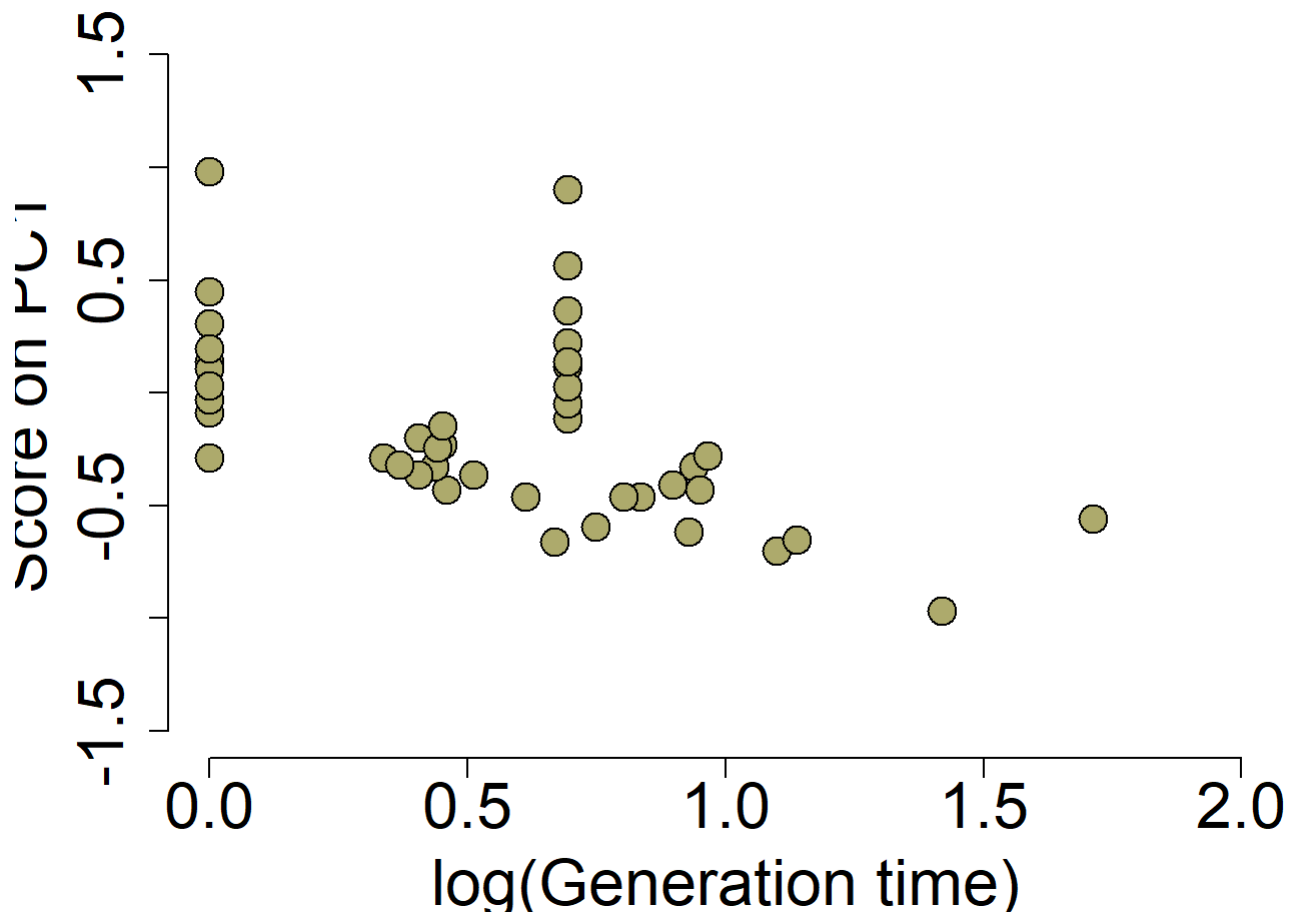
```
# Covariances
test <- as.matrix(scale(log(data_squirrels[,c(2,3,4,7)])))
cov(test)
```

```
##              age_first_repro lifetime_adult breed_freq   fec_one
## age_first_repro      1.00000000 -0.03796442      NA 0.08742549
## lifetime_adult      -0.03796442  1.00000000      NA 0.34347277
## breed_freq              NA              NA      NA      NA
## fec_one              0.08742549  0.34347277      NA 1.00000000
```

```
# Correlation with generation time
individual_score_squirrels_reverse <- as.numeric(vegan::scores(test.pca_squirrels1_reverse)$sites[,1])
summary(lm(individual_score_squirrels_reverse ~ log(data_squirrels$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_squirrels_reverse ~ log(data_squirrels$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.5871 -0.2682 -0.1022  0.1847  1.0604
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.29934    0.07053   4.244 7.46e-05 ***
## log(data_squirrels$generation_time) -0.66021    0.11489  -5.746 2.97e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3804 on 62 degrees of freedom
## Multiple R-squared:  0.3475, Adjusted R-squared:  0.337
## F-statistic: 33.02 on 1 and 62 DF, p-value: 2.974e-07
```

```
par(bty="n")
plot(log(data_squirrels$generation_time), individual_score_squirrels_reverse, pch=21, xlab="log
(Generation time)", ylab="Score on PC1", xlim=c(0,2), ylim=c(-1.5,1.5), bg=colors[6], cex=2, ce
x.axis=2, cex.lab=2)
```



Bighorn sheep

```
data_sheep <- data[which(data$species == "bighornsheep"),]  
data_sheep$fec_one <- data_sheep$fecundity + 1  
test.pca_sheep1 <- rda(log(data_sheep[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_sheep1)
```

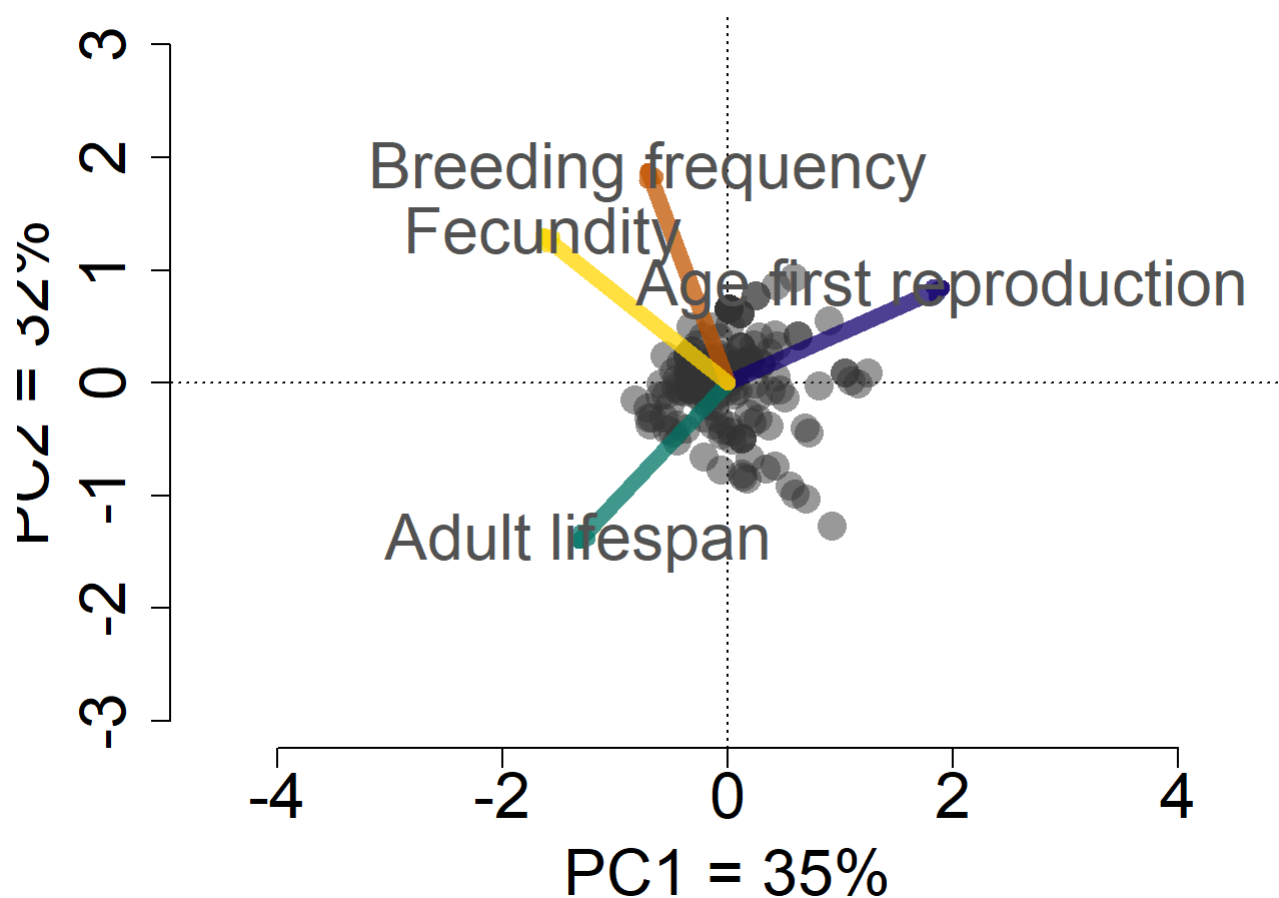
```
##
## Call:
## rda(X = log(data_sheep[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained    4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      1.3805 1.2769 0.7816 0.5610
## Proportion Explained 0.3451 0.3192 0.1954 0.1403
## Cumulative Proportion 0.3451 0.6643 0.8597 1.0000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 4.981902
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro 1.903 0.8447 0.7872 -1.1187
## lifetime_adult -1.325 -1.4038 1.4292 -0.6598
## breed_freq      -0.706 1.8808 1.2554 0.7701
## fec_one         -1.640 1.3045 -0.7818 -1.0959
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 1  -0.340652 0.020521 0.224293 -0.151580
## 2  -0.358280 0.243806 -0.075674 -0.224049
## 3   0.040524 -0.508740 0.277606 0.134585
## 4  -0.702574 -0.311992 0.219438 0.336625
## 5  -0.693030 -0.372255 -0.157806 0.041883
## 6   0.128481 -0.500661 -0.098662 0.025497
## 7  -0.304921 0.032731 0.195562 -0.101809
## 8  -0.345234 -0.049825 0.322202 -0.143591
## 9   0.172688 -0.849951 -0.675043 -0.568663
## 10 -0.331661 0.498793 -1.015902 0.710358
## 11 -0.085154 -0.108216 0.314821 0.249457
## 12 -0.344779 0.062452 -0.413732 0.782321
## 13 -0.007264 0.224450 -0.167082 0.294004
## 14 -0.567389 -0.310998 0.172662 0.534374
## 15 -0.321227 0.286237 -0.146250 -0.178656
## 16 -0.211929 0.055180 0.133567 0.029673
## 17 0.459983 -0.064564 0.484792 0.365488
```

## 18	0.691167	-1.029865	-0.583406	-0.721908
## 19	-0.389159	-0.074870	0.371269	-0.202679
## 20	-0.215471	-0.309942	0.138605	-0.248505
## 21	0.295183	0.168034	0.221510	0.075562
## 22	-0.148131	0.063139	0.101231	0.121433
## 23	0.335361	-0.762373	-1.243233	-0.625706
## 24	-0.366292	0.085492	0.143900	-0.202702
## 25	0.055713	0.053095	0.046508	0.422028
## 26	0.804796	-0.026069	-0.096697	1.535486
## 27	-0.121966	0.205847	-0.103063	0.129928
## 28	-0.330167	-0.062779	0.334886	-0.118820
## 29	0.232206	0.339389	0.007920	-0.052463
## 30	0.175445	0.180057	0.245266	-0.102288
## 31	-0.542907	-0.415928	-0.363698	0.119343
## 32	-0.711612	-0.219621	0.095931	0.304089
## 33	0.005854	0.660790	-0.769252	0.222041
## 34	-0.148131	0.063139	0.101231	0.121433
## 35	-0.274769	-0.036119	0.279753	-0.043270
## 36	0.417049	-0.734036	-0.036594	0.383740
## 37	-0.081669	-0.319993	-0.275567	-0.319984
## 38	-0.340652	0.020521	0.224293	-0.151580
## 39	-0.451918	-0.374818	0.221299	0.716799
## 40	0.117504	0.320787	0.071939	-0.216539
## 41	1.158513	-0.004805	0.750220	0.427185
## 42	-0.358280	0.243806	-0.075674	-0.224049
## 43	0.361193	0.269210	0.381007	-0.371028
## 44	-0.554928	0.243120	-0.590637	0.436839
## 45	-0.822781	-0.154242	0.043714	0.127638
## 46	0.052834	-0.109633	0.270404	0.451815
## 47	0.245324	0.775730	-0.594250	-0.124425
## 48	0.010364	0.001164	0.132886	0.366473
## 49	0.127376	-0.485001	-0.280442	-0.092476
## 50	0.504865	-0.136722	0.888846	-0.075824
## 51	0.104458	0.614418	-0.325938	-0.296996
## 52	-0.121966	0.205847	-0.103063	0.129928
## 53	-0.051548	-0.465037	0.248666	-0.009372
## 54	-0.204891	-0.096194	0.338577	0.071608
## 55	-0.135230	-0.374802	0.095832	-0.191109
## 56	0.055713	0.053095	0.046508	0.422028
## 57	-0.262832	0.044536	0.165250	-0.042643
## 58	-0.229875	-0.149001	0.419312	0.046058
## 59	0.267153	-0.318267	0.135414	-0.399896
## 60	0.688354	-0.397623	0.144585	-0.484496
## 61	-0.167555	-0.202578	0.471773	0.148510
## 62	-0.121966	0.205847	-0.103063	0.129928
## 63	-0.262832	0.044536	0.165250	-0.042643
## 64	-0.595502	-0.004413	-0.237898	0.429146
## 65	-0.345234	-0.049825	0.322202	-0.143591
## 66	0.196579	0.214406	-0.221804	0.594599
## 67	-0.453374	-0.514619	-0.187713	0.320937
## 68	0.436049	0.329346	-0.046803	0.248133
## 69	1.089148	0.016712	0.482358	0.747709

## 70	0.005854	0.660790	-0.769252	0.222041
## 71	0.621797	0.418499	0.088939	-0.020607
## 72	-0.481463	0.102745	0.158957	-0.374957
## 73	-0.059067	-0.772288	-0.639071	0.479372
## 74	-0.081757	0.401176	0.028752	-0.525123
## 75	0.555407	-0.913150	-0.697689	-0.945095
## 76	0.060012	0.142203	0.335903	-0.263412
## 77	-0.204891	-0.096194	0.338577	0.071608
## 78	0.027541	0.170119	0.308569	-0.316793
## 79	-0.230532	-0.031222	0.258183	0.020485
## 80	-0.340652	0.020521	0.224293	-0.151580
## 81	0.027683	-0.433206	-0.039587	-0.053430
## 82	0.192483	-0.665804	0.223695	0.236496
## 83	0.194686	0.172870	0.568926	-0.594721
## 84	0.592563	-0.983493	-1.026719	-0.202871
## 85	0.187990	-0.271311	0.187518	-0.462412
## 86	0.211324	-0.356047	0.471297	-0.287014
## 87	-0.216384	-0.656313	-1.113255	-0.034300
## 88	-0.367725	-0.408367	0.238974	0.847097
## 89	-0.148131	0.063139	0.101231	0.121433
## 90	-0.321227	0.286237	-0.146250	-0.178656
## 91	0.237691	-0.010549	0.485474	0.028688
## 92	-0.041471	-0.416574	0.121559	-0.045086
## 93	0.196579	0.214406	-0.221804	0.594599
## 94	0.430943	0.051319	0.335793	0.298749
## 95	0.128481	-0.500661	-0.098662	0.025497
## 96	-0.459076	-0.042871	0.350922	-0.311748
## 97	-0.416745	0.176856	0.035686	-0.295672
## 98	-0.199786	0.181832	-0.044019	0.020991
## 99	1.044267	0.088870	0.078304	1.189020
## 100	-0.288997	-0.098173	0.369543	-0.051138
## 101	0.104458	0.614418	-0.325938	-0.296996
## 102	0.175445	0.180057	0.245266	-0.102288
## 103	0.005934	-0.380597	-0.382672	-0.292124
## 104	-0.600347	-0.117462	-0.081401	0.445674
## 105	0.431072	0.864883	-0.458509	-0.393165
## 106	0.125363	-0.809196	-0.229201	-0.304636
## 107	0.295183	0.168034	0.221510	0.075562
## 108	-0.423053	-0.073840	0.381246	-0.252527
## 109	-0.671860	-0.312101	0.209270	0.381623
## 110	0.134006	-0.835790	-0.403128	-0.432411
## 111	-0.262832	0.044536	0.165250	-0.042643
## 112	-0.487510	-0.442451	-0.435809	0.142787
## 113	0.272796	0.411433	0.477555	-0.956783
## 114	-0.064718	0.249435	0.230904	-0.468466
## 115	-0.549443	-0.106818	-0.113083	0.517990
## 116	-0.342356	0.112903	0.098307	-0.173378
## 117	-0.514928	-0.190897	-0.009495	0.586101
## 118	-0.114404	-0.212821	0.467949	0.228482
## 119	0.360601	-0.379376	-0.691603	-0.869083
## 120	0.582839	0.937727	-0.347600	-0.612742
## 121	0.117504	0.320787	0.071939	-0.216539

```
## 122 0.091340 0.178078 0.276232 -0.225034
## 123 0.385203 -0.072681 0.841313 -0.264431
## 124 -0.013033 -0.077415 0.661307 -0.324486
## 125 0.055189 -0.037807 0.584127 -0.232862
## 126 1.240915 0.089556 0.593267 0.528132
## 127 0.417953 0.428543 0.143661 -0.321203
## 128 0.062057 0.086039 0.412157 -0.248682
## 129 0.724027 -0.442974 -0.406291 -0.845723
## 130 -0.102885 0.227842 0.273308 -0.519844
## 131 0.901881 0.552932 0.293620 -0.425834
## 132 -0.101182 0.135460 0.399295 -0.498046
## 133 -0.211929 0.055180 0.133567 0.029673
## 134 0.265862 -0.127573 0.223413 0.767509
## 135 -0.065451 0.147670 0.370564 -0.448275
## 136 0.245324 0.775730 -0.594250 -0.124425
## 137 -0.450091 0.040321 0.233933 -0.315973
## 138 0.117504 0.320787 0.071939 -0.216539
## 139 -0.342356 0.112903 0.098307 -0.173378
## 140 -0.304921 0.032731 0.195562 -0.101809
## 141 0.022057 0.520057 -0.168985 -0.397944
## 142 -0.358280 0.243806 -0.075674 -0.224049
## 143 -0.275879 0.338168 -0.232627 -0.123101
## 144 -0.135012 0.499479 -0.500939 0.049470
## 145 0.005854 0.660790 -0.769252 0.222041
## 146 0.005854 0.660790 -0.769252 0.222041
## 147 -0.217414 0.405118 -0.343987 -0.051478
## 148 0.005854 0.660790 -0.769252 0.222041
## 149 0.089887 -0.067203 0.199829 0.497207
## 150 1.044267 0.088870 0.078304 1.189020
## 151 0.621797 0.418499 0.088939 -0.020607
## 152 0.924154 -1.268565 -0.747584 0.342258
## 153 0.104458 0.614418 -0.325938 -0.296996
## 154 0.005854 0.660790 -0.769252 0.222041
## 155 0.005854 0.660790 -0.769252 0.222041
```

```
# Figure
par(mar=c(4,4,2,2), bty="n")
plot(test.pca_sheep1, scaling = scl, choices=c(1,2), type="n", main="", ylim=c(-3,3), xlim=c(-3.5,3.5), xlab="PC1 = 35%", ylab="PC2 = 32%", cex.lab=2, cex.axis=2)
with(data_sheep, points(test.pca_sheep1, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_sheep1, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```

# Test for proportion of variance explained per axe
sum_all_sheep <- summary(test.pca_sheep1)

prop_sheep <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_sheep[j,i+1] <- as.numeric(abs(sum_all_sheep$species[i,j])/sum(abs(sum_all_sheep$species[i,])))
  }
}

prop_sheep

```

```

##   PC      prim      adult      breed      freq
## 1  1 0.4088800 0.2750622 0.1530603 0.3401417
## 2  2 0.1815314 0.2913592 0.4077764 0.2704904
## 3  3 0.1691780 0.2966354 0.2721918 0.1621174
## 4  4 0.2404107 0.1369432 0.1669715 0.2272505

```

```

# Covariances
test <- as.matrix(scale(log(data_sheep[,c(2,3,4,7)])))
cov(test)

```

```

##           age_first_repro lifetime_adult breed_freq      fec_one
## age_first_repro      1.00000000 -0.297196747 0.06000607 -0.227013494
## lifetime_adult      -0.29719675  1.000000000 -0.06744570 -0.008310028
## breed_freq           0.06000607 -0.067445704 1.00000000 0.287826653
## fec_one              -0.22701349 -0.008310028 0.28782665 1.000000000

```

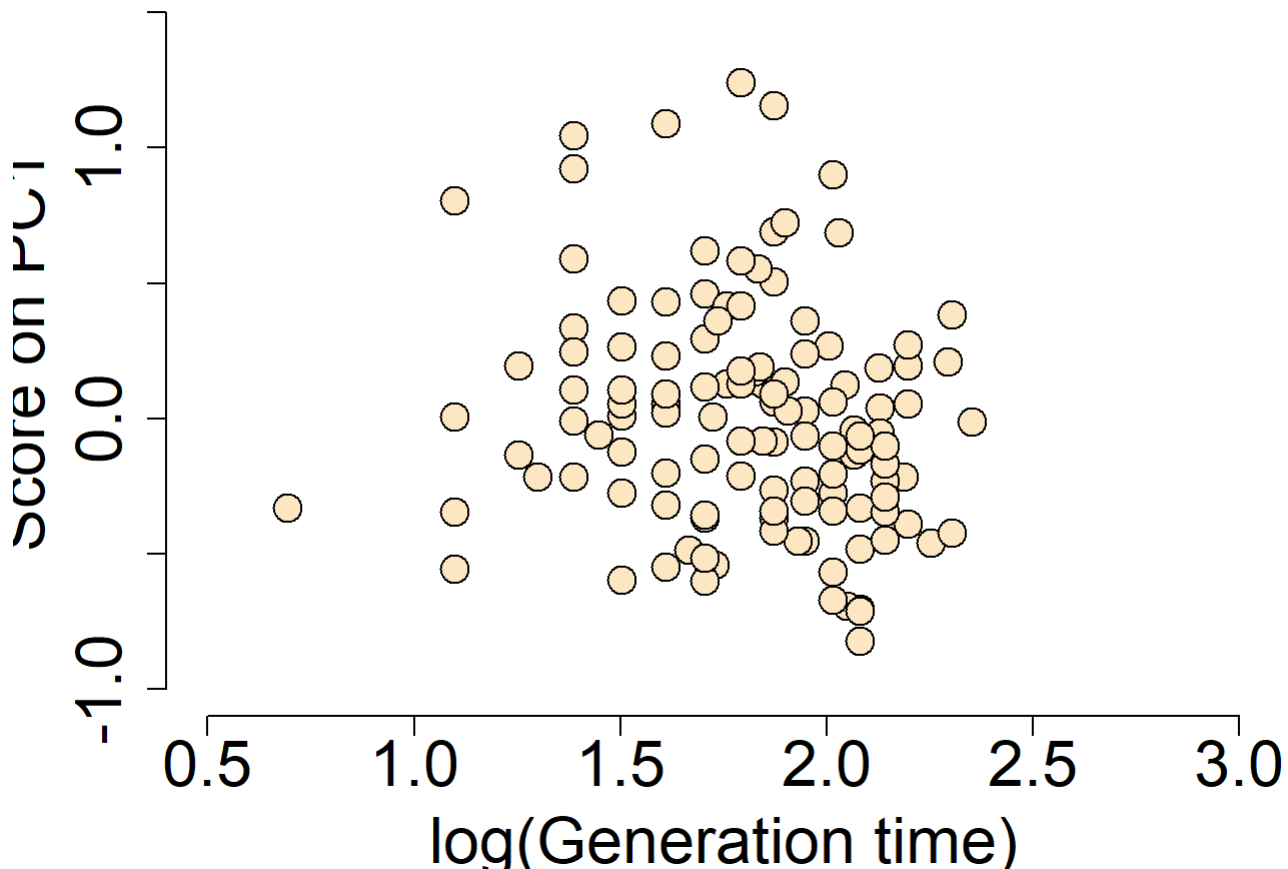
```

# Correlation with generation time
individual_score_sheep <- as.numeric(vegan::scores(test.pca_sheep1)$sites[,1])
summary(lm(individual_score_sheep ~ log(data_sheep$generation_time)))

```

```
##
## Call:
## lm(formula = individual_score_sheep ~ log(data_sheep$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.74895 -0.26196 -0.04879  0.18920  1.24825
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.4068     0.1805   2.254  0.0256 *
## log(data_sheep$generation_time) -0.2311     0.1009  -2.290  0.0234 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.396 on 153 degrees of freedom
## Multiple R-squared:  0.03313,    Adjusted R-squared:  0.02681
## F-statistic: 5.243 on 1 and 153 DF,  p-value: 0.0234
```

```
par(bty="n")
plot(log(data_sheep$generation_time), individual_score_sheep, pch=21, xlab="log(Generation time)", ylab="Score on PC1", xlim=c(0.5,3), ylim=c(-1,1.5), bg=colors[2], cex=2, cex.lab=2, cex.axis=2)
```



###

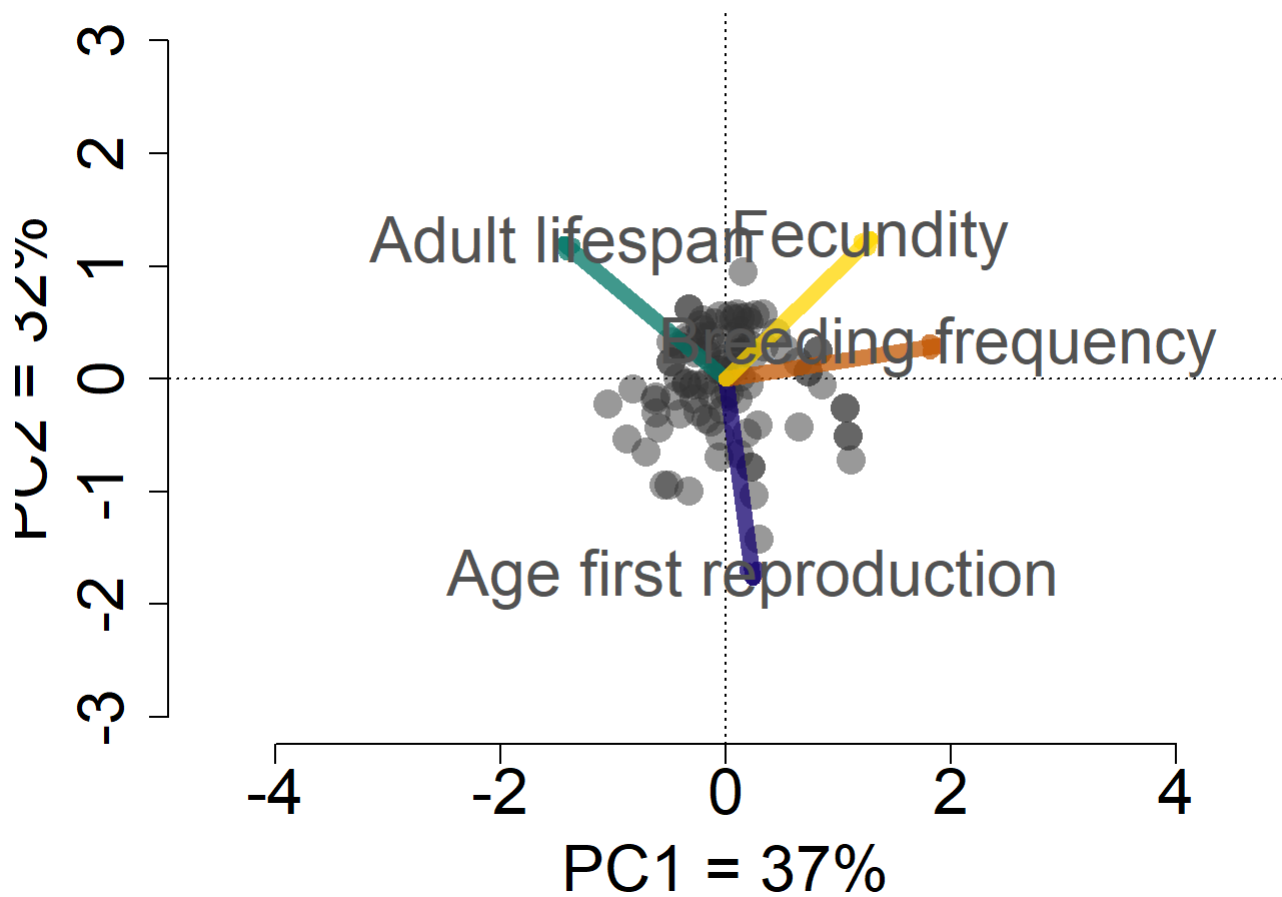
```
data_goats <- data[which(data$species == "mountaingoat"),]  
data_goats$fec_one <- data_goats$fecundity + 1  
test.pca_goats1 <- rda(log(data_goats[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_goats1)
```

```
##
## Call:
## rda(X = log(data_goats[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained   4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      1.4938 1.2603 0.7455 0.5004
## Proportion Explained 0.3735 0.3151 0.1864 0.1251
## Cumulative Proportion 0.3735 0.6885 0.8749 1.0000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 4.426728
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro 0.2423 -1.7623 1.2768 -0.3231
## lifetime_adult -1.4406 1.1984 0.7831 -0.8801
## breed_freq      1.8822 0.2983 -0.3392 -1.0734
## fec_one         1.2813 1.2423 1.1373 0.6484
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 156 0.128363 0.426638 -0.080740 -0.083423
## 157 0.201199 0.510349 0.048821 0.026620
## 158 -0.010280 0.062041 0.148658 -0.583623
## 159 -0.062487 -0.690739 0.012590 0.256342
## 160 1.111910 -0.716142 0.403883 0.548206
## 161 0.445390 0.410507 -0.173655 -0.124889
## 162 -0.113845 0.316379 0.587766 -0.313243
## 163 -0.052073 0.558360 0.097755 -0.231837
## 164 0.103432 0.229973 0.374310 -0.522582
## 165 -0.101272 0.501816 0.010240 -0.306169
## 166 -0.203795 0.438248 -0.051351 -0.279713
## 167 0.280675 -0.403860 0.095127 -0.725035
## 168 -0.039614 0.084932 0.364438 0.122338
## 169 -0.172413 0.224968 0.475161 -0.303562
## 170 0.161873 0.243022 0.417326 -0.379923
## 171 -0.331096 0.625571 -0.578501 0.763331
## 172 -0.283047 -0.173731 0.655151 -0.189803
```

## 173	-0.022504	0.137511	0.266653	-0.538137
## 174	-0.623692	-0.156251	0.140847	0.116258
## 175	-0.325740	0.349988	-0.098924	-0.077728
## 176	0.332969	0.582086	0.030368	-0.338820
## 177	0.521830	0.271042	0.428200	-0.219831
## 178	0.257926	0.571216	-0.001875	-0.463242
## 179	0.236406	0.206816	0.420615	-0.113518
## 180	0.851085	-0.053617	0.069549	0.380653
## 181	0.137948	0.366662	0.548665	-0.553240
## 182	-0.207620	0.528501	0.103368	-0.163424
## 183	-0.488522	0.326016	-0.079255	0.008694
## 184	-0.232533	0.477154	0.057647	-0.082712
## 185	-0.280296	-0.054665	0.039813	-0.579683
## 186	0.076594	0.094166	-0.301791	0.817758
## 187	1.058822	-0.258453	-0.156734	0.759516
## 188	-0.388522	0.277833	-0.210599	-0.172579
## 189	0.185641	0.555566	-0.039891	-0.582334
## 190	0.217039	0.306425	0.515455	-0.296576
## 191	0.095089	0.577691	0.079928	-0.297374
## 192	0.039660	0.561672	0.033631	-0.434540
## 193	0.068225	0.533506	0.002517	-0.382445
## 194	-0.454197	-0.152466	0.159140	0.295028
## 195	-0.596625	-0.436391	0.374420	-0.101687
## 196	0.122360	0.543832	-0.070530	-0.686880
## 197	0.114055	-0.667567	0.506587	-0.059238
## 198	-0.347672	-0.034366	0.197024	-0.119598
## 199	0.338212	0.287327	-0.364304	-0.286818
## 200	0.147866	0.955639	-0.505671	-0.390160
## 201	-0.489404	0.151067	-0.212653	0.435166
## 202	-0.822505	-0.088696	-0.486722	0.410070
## 203	0.025582	-0.124352	0.163918	0.520379
## 204	0.166033	0.254868	-0.176309	0.504136
## 205	-0.143654	0.378946	-0.116861	-0.170030
## 206	-0.177472	-0.357720	-0.197272	0.213600
## 207	0.185137	-0.474993	-1.424203	-0.432150
## 208	-0.347020	0.377806	-0.020420	0.064835
## 209	-0.229317	0.371733	-0.066152	0.005488
## 210	0.729567	0.066205	0.201916	0.159032
## 211	-0.349953	-0.043216	0.245637	0.119108
## 212	-0.408551	-0.305906	-0.293402	-0.622297
## 213	-0.273217	0.234368	-0.195725	0.209789
## 214	0.108487	-0.176847	0.631629	0.027877
## 215	-0.146770	-0.005042	0.172253	-0.226791
## 216	-0.420460	0.295123	-0.087705	0.194867
## 217	0.044514	0.374690	-0.043941	0.282515
## 218	-0.054270	-0.507746	-0.753279	-1.073118
## 219	0.249110	-1.026530	-0.748632	-0.686789
## 220	-0.331096	0.625571	-0.578501	0.763331
## 221	-0.073933	0.078668	0.301814	-0.116748
## 222	-0.009625	0.179181	-0.207875	0.660516
## 223	-0.010848	-0.163072	0.668216	0.007838
## 224	-0.503712	-0.943139	0.374673	0.459351

```
## 225 0.815879 0.249916 -0.302244 0.520790
## 226 -0.709923 -0.645636 0.626135 -0.324359
## 227 0.220343 -0.778525 -1.052410 -0.572288
## 228 -0.212679 -0.054188 -0.569065 0.353738
## 229 -0.619590 -0.298133 0.050888 0.537768
## 230 -0.881925 -0.531432 0.451761 0.508480
## 231 1.058822 -0.258453 -0.156734 0.759516
## 232 -0.625391 -0.188387 0.070642 0.012814
## 233 0.294499 -1.417851 -0.269307 -0.867457
## 234 0.729567 0.066205 0.201916 0.159032
## 235 -1.044399 -0.220375 0.266486 0.647098
## 236 0.643348 0.151220 0.295832 0.001790
## 237 -0.022600 -0.270156 -1.197920 -0.811013
## 238 -0.129068 -0.377204 0.950819 -0.069794
## 239 -0.171849 0.252119 -1.142306 0.691649
## 240 -0.426099 0.008758 -0.262624 1.061454
## 241 -0.545434 -0.937458 0.833990 0.242598
## 242 0.201240 -0.048665 0.195485 0.363998
## 243 0.220343 -0.778525 -1.052410 -0.572288
## 244 0.650628 -0.422163 1.151782 -0.415346
## 245 1.087589 -0.506458 0.147043 0.645015
## 246 -0.324268 -0.991604 -0.453065 -0.139164
## 247 0.134363 0.017278 0.268332 0.242032
## 248 -0.126460 -0.139203 -0.662981 0.510980
## 249 -0.244348 -0.291777 -0.124425 0.091634
## 250 -0.489404 0.151067 -0.212653 0.435166
## 251 0.815879 0.249916 -0.302244 0.520790
## 252 1.087589 -0.506458 0.147043 0.645015
```

```
# Figure
par(mar=c(4,4,2,2), bty="n")
plot(test.pca_goats1, scaling = scl, choices=c(1,2), type="n", main="", ylim=c(-3,3), xlim=c(-3.5,3.5), xlab="PC1 = 37%", ylab="PC2 = 32%", cex.axis=2, cex.lab=2)
with(data_goats, points(test.pca_goats1, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_goats1, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```
# Test for proportion of variance explained per axe
sum_all_goats <- summary(test.pca_goats1)

prop_goats <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_goats[j,i+1] <- as.numeric(abs(sum_all_goats$species[i,j])/sum(abs(sum_all_goats$species[i,])))
  }
}

prop_goats
```

```
##      PC      prim      adult      breed      freq
## 1  1 0.06721900 0.3348530 0.52383179 0.2973203
## 2  2 0.48892196 0.2785597 0.08303177 0.2882878
## 3  3 0.35423277 0.1820186 0.09441277 0.2639201
## 4  4 0.08962628 0.2045687 0.29872367 0.1504719
```

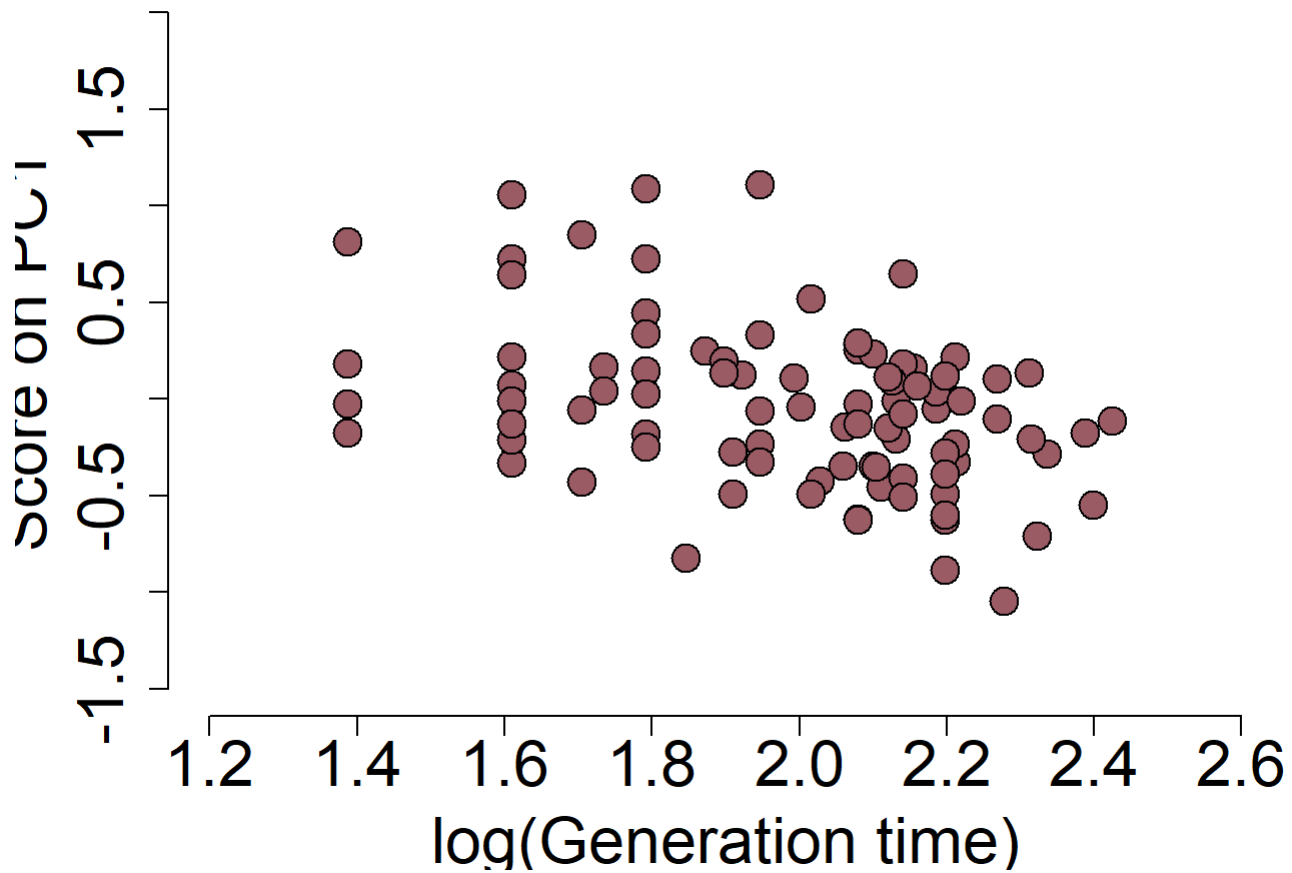
```
# Covariances
test <- as.matrix(scale(log(data_goats[,c(2,3,4,7)])))
cov(test)
```

```
##      age_first_repro lifetime_adult breed_freq fec_one
## age_first_repro      1.00000000 -0.240212751 -0.03187032 -0.129874036
## lifetime_adult      -0.24021275  1.000000000 -0.34189148 -0.007556062
## breed_freq          -0.03187032 -0.341891477 1.00000000  0.347096568
## fec_one             -0.12987404 -0.007556062 0.34709657  1.000000000
```

```
# Correlation with generation time
individual_score_goats <- as.numeric(vegan::scores(test.pca_goats1)$sites[,1])
summary(lm(individual_score_goats ~ log(data_goats$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_goats ~ log(data_goats$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.91135 -0.30689 -0.01168  0.29717  1.09815
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.4738     0.3195   4.613 1.24e-05 ***
## log(data_goats$generation_time) -0.7503     0.1613  -4.652 1.06e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4099 on 95 degrees of freedom
## Multiple R-squared:  0.1856, Adjusted R-squared:  0.177
## F-statistic: 21.64 on 1 and 95 DF, p-value: 1.061e-05
```

```
par(bty="n")
plot(log(data_goats$generation_time), individual_score_goats, pch=21, xlab="log(Generation time)", ylab="Score on PC1", xlim=c(1.2,2.6), ylim=c(-1.5,2), bg=colors[10], cex=2, cex.axis=2, cex.lab=2)
```



White-throated dipper

```
data_dipper <- data[which(data$species == "dipper"),]
data_dipper$fec_one <- data_dipper$fecundity + 1
test.pca_dipper1 <- rda(log(data_dipper[,c(2,3,4,7)]), scale=TRUE)
summary(test.pca_dipper1)
```

```
##
## Call:
## rda(X = log(data_dipper[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained   4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      1.3483 0.9836 0.9713 0.6968
## Proportion Explained 0.3371 0.2459 0.2428 0.1742
## Cumulative Proportion 0.3371 0.5830 0.8258 1.0000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 8.472156
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro 1.441 -1.9917 3.407 0.5415
## lifetime_adult  3.199 0.5414 -1.204 2.4427
## breed_freq      -1.839 -3.2584 -1.586 1.1961
## fec_one         2.915 -1.6653 -1.364 -2.1939
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 1295 -0.062345 -0.0585139 -0.020285 -0.205078
## 2102 0.318866 -0.0079075 -0.224086 0.248779
## 3102 -0.293864 0.1227631 0.130031 0.132033
## 4102 0.005370 -0.1115340 -0.064250 -0.303676
## 5102 0.401867 0.0292056 -0.265193 0.423015
## 6102 -0.025079 -0.0876930 -0.044481 -0.259341
## 7102 0.061582 0.0189972 -0.078894 0.118939
## 8102 0.031134 0.0428383 -0.059125 0.163275
## 9102 -0.025079 -0.0876930 -0.044481 -0.259341
## 10102 0.005370 -0.1115340 -0.064250 -0.303676
## 11102 -0.293864 0.1227631 0.130031 0.132033
## 1296 -0.293864 0.1227631 0.130031 0.132033
## 1313 -0.293864 0.1227631 0.130031 0.132033
## 1413 0.129298 -0.0340229 -0.122859 0.020340
## 1513 0.061582 0.0189972 -0.078894 0.118939
## 1613 1.182023 2.4114256 1.027155 -1.022407
## 1713 -0.110390 -0.0208955 0.010908 -0.135121
```

## 1813	0.235968	-0.0154427	-0.179334	0.160112
## 1912	0.162811	-0.0602634	-0.144618	-0.028458
## 2012	0.258269	-0.0329038	-0.193813	0.127641
## 2112	0.091410	-0.6071076	1.394358	0.412161
## 2212	-0.062345	-0.0585139	-0.020285	-0.205078
## 2312	0.162811	-0.0602634	-0.144618	-0.028458
## 2412	0.210224	0.0047146	-0.162619	0.197597
## 2512	0.210224	0.0047146	-0.162619	0.197597
## 2612	-0.293864	0.1227631	0.130031	0.132033
## 2712	-0.293864	0.1227631	0.130031	0.132033
## 2812	-0.025079	-0.0876930	-0.044481	-0.259341
## 2912	-0.025079	-0.0876930	-0.044481	-0.259341
## 3012	-0.025079	-0.0876930	-0.044481	-0.259341
## 3112	0.005370	-0.1115340	-0.064250	-0.303676
## 3212	0.146893	-0.0478003	-0.134283	-0.005281
## 3312	-0.025079	-0.0876930	-0.044481	-0.259341
## 3412	0.291724	0.0133441	-0.206464	0.288299
## 3512	-0.062345	-0.0585139	-0.020285	-0.205078
## 3612	-0.293864	0.1227631	0.130031	0.132033
## 3712	0.327014	-0.0142875	-0.229376	0.236914
## 3812	-0.293864	0.1227631	0.130031	0.132033
## 3912	-0.062345	-0.0585139	-0.020285	-0.205078
## 4012	0.146893	-0.0478003	-0.134283	-0.005281
## 4112	0.342207	-0.0261833	-0.239241	0.214792
## 4212	0.177342	-0.0716413	-0.154053	-0.049617
## 4312	0.005370	-0.1115340	-0.064250	-0.303676
## 4412	0.005370	-0.1115340	-0.064250	-0.303676
## 4512	0.430241	0.0069890	-0.283615	0.381700
## 4612	-0.062345	-0.0585139	-0.020285	-0.205078
## 4712	0.177342	-0.0716413	-0.154053	-0.049617
## 4812	0.414369	0.0582347	-0.268450	0.517002
## 4912	0.129298	-0.0340229	-0.122859	0.020340
## 5012	0.162811	-0.0602634	-0.144618	-0.028458
## 5112	0.005370	-0.1115340	-0.064250	-0.303676
## 5212	0.246922	0.0484233	-0.177376	0.353534
## 5312	0.162811	-0.0602634	-0.144618	-0.028458
## 5412	-0.293864	0.1227631	0.130031	0.132033
## 5512	0.413801	0.0198613	-0.272941	0.405638
## 5612	0.235968	-0.0154427	-0.179334	0.160112
## 5712	0.162811	-0.0602634	-0.144618	-0.028458
## 5812	0.129298	-0.0340229	-0.122859	0.020340
## 5912	0.087327	-0.0011601	-0.095609	0.081453
## 6012	0.387081	-0.0051286	-0.261341	0.311853
## 6112	-0.293864	0.1227631	0.130031	0.132033
## 6212	-0.062345	-0.0585139	-0.020285	-0.205078
## 6312	-0.293864	0.1227631	0.130031	0.132033
## 6412	-0.293864	0.1227631	0.130031	0.132033
## 6511	0.263408	-0.0369276	-0.197149	0.120158
## 6611	0.532168	-0.7776709	1.130043	0.274847
## 6711	-0.025079	-0.0876930	-0.044481	-0.259341
## 6811	0.031114	-0.1316914	-0.080965	-0.341162
## 6911	-0.121892	0.1626558	0.040229	0.386093

## 7011	0.562617	-0.8015120	1.110274	0.230511
## 7111	0.416388	-0.8615620	1.183362	-0.061035
## 7211	-0.025079	-0.0876930	-0.044481	-0.259341
## 7311	0.291724	0.0133441	-0.206464	0.288299
## 7411	0.005370	-0.1115340	-0.064250	-0.303676
## 7511	0.418045	-0.0293726	-0.281444	0.266768
## 7611	1.629873	1.6325586	2.250853	-0.833395
## 7711	0.258269	-0.0329038	-0.193813	0.127641
## 7811	0.195693	0.0160925	-0.153184	0.218756
## 7911	0.210224	0.0047146	-0.162619	0.197597
## 8011	0.334783	-0.0203706	-0.234421	0.225602
## 8111	0.342207	-0.0261833	-0.239241	0.214792
## 8211	-0.293864	0.1227631	0.130031	0.132033
## 8311	0.339352	-0.0963906	-0.246456	0.009577
## 8411	0.005370	-0.1115340	-0.064250	-0.303676
## 8511	-0.025079	-0.0876930	-0.044481	-0.259341
## 8611	0.258269	-0.0329038	-0.193813	0.127641
## 8710	0.268394	-0.0408313	-0.200386	0.112898
## 8811	0.319366	0.0478915	-0.217376	0.410452
## 8911	-0.110390	-0.0208955	0.010908	-0.135121
## 9011	0.005370	-0.1115340	-0.064250	-0.303676
## 9111	0.548085	-0.7901340	1.119708	0.251670
## 9210	0.399015	-0.0144729	-0.269089	0.294476
## 9310	0.031114	-0.1316914	-0.080965	-0.341162
## 9410	-0.025079	-0.0876930	-0.044481	-0.259341
## 9510	0.223592	-0.0057521	-0.171298	0.178133
## 9610	-0.110390	-0.0208955	0.010908	-0.135121
## 9711	-0.293864	0.1227631	0.130031	0.132033
## 9811	-0.293864	0.1227631	0.130031	0.132033
## 9911	-0.062345	-0.0585139	-0.020285	-0.205078
## 10011	-0.293864	0.1227631	0.130031	0.132033
## 10111	-0.110390	-0.0208955	0.010908	-0.135121
## 10211	0.430241	0.0069890	-0.283615	0.381700
## 10311	0.327014	-0.0142875	-0.229376	0.236914
## 10411	-0.293864	0.1227631	0.130031	0.132033
## 10511	0.162811	-0.0602634	-0.144618	-0.028458
## 10611	0.203086	-0.0917986	-0.170767	-0.087102
## 10711	0.087327	-0.0011601	-0.095609	0.081453
## 10811	0.146893	-0.0478003	-0.134283	-0.005281
## 10911	-0.110390	-0.0208955	0.010908	-0.135121
## 11011	0.146893	-0.0478003	-0.134283	-0.005281
## 11111	-0.293864	0.1227631	0.130031	0.132033
## 11211	-0.121892	0.1626558	0.040229	0.386093
## 11310	-0.121892	0.1626558	0.040229	0.386093
## 11411	0.061582	0.0189972	-0.078894	0.118939
## 11511	0.005370	-0.1115340	-0.064250	-0.303676
## 11611	-0.293864	0.1227631	0.130031	0.132033
## 11711	-0.293864	0.1227631	0.130031	0.132033
## 11811	-0.293864	0.1227631	0.130031	0.132033
## 11911	-0.293864	0.1227631	0.130031	0.132033
## 12011	0.146893	-0.0478003	-0.134283	-0.005281
## 12111	0.380778	-0.0001936	-0.257249	0.321030

##	12211	0.223592	-0.0057521	-0.171298	0.178133
##	12311	0.109627	-0.0186212	-0.110088	0.048982
##	12410	0.442743	0.0360181	-0.286872	0.475687
##	12511	0.005370	-0.1115340	-0.064250	-0.303676
##	12611	0.005370	-0.1115340	-0.064250	-0.303676
##	12711	0.177342	-0.0716413	-0.154053	-0.049617
##	12811	0.162811	-0.0602634	-0.144618	-0.028458
##	1297	0.374228	0.0049351	-0.252996	0.330568
##	1302	-0.293864	0.1227631	0.130031	0.132033
##	1314	-0.062345	-0.0585139	-0.020285	-0.205078
##	1322	-0.062345	-0.0585139	-0.020285	-0.205078
##	1332	0.177342	-0.0716413	-0.154053	-0.049617
##	1342	0.087327	-0.0011601	-0.095609	0.081453
##	1352	0.360195	-0.8175637	1.219846	0.020787
##	1362	0.647927	-0.8683094	1.054885	0.106291
##	1372	0.005370	-0.1115340	-0.064250	-0.303676
##	1382	0.258269	-0.0329038	-0.193813	0.127641
##	1392	-0.293864	0.1227631	0.130031	0.132033
##	1402	0.005370	-0.1115340	-0.064250	-0.303676
##	1414	0.162180	0.0423330	-0.131426	0.267554
##	1422	0.162811	-0.0602634	-0.144618	-0.028458
##	1432	0.784290	-0.7443435	0.995237	0.574603
##	1442	0.595499	-0.7251561	1.101707	0.477725
##	1452	0.061582	0.0189972	-0.078894	0.118939
##	1462	0.005370	-0.1115340	-0.064250	-0.303676
##	1472	0.195693	0.0160925	-0.153184	0.218756
##	1482	0.005370	-0.1115340	-0.064250	-0.303676
##	1492	-0.006133	0.0720173	-0.034930	0.217538
##	1502	-0.025079	-0.0876930	-0.044481	-0.259341
##	1514	0.185510	0.0965083	-0.137503	0.442956
##	1522	0.195693	0.0160925	-0.153184	0.218756
##	1532	-0.293864	0.1227631	0.130031	0.132033
##	1542	0.291724	0.0133441	-0.206464	0.288299
##	1552	-0.293864	0.1227631	0.130031	0.132033
##	1562	-0.110390	-0.0208955	0.010908	-0.135121
##	1572	-0.293864	0.1227631	0.130031	0.132033
##	1582	0.258269	-0.0329038	-0.193813	0.127641
##	1591	-0.025079	-0.0876930	-0.044481	-0.259341
##	1602	0.129298	-0.0340229	-0.122859	0.020340
##	1614	0.087327	-0.0011601	-0.095609	0.081453
##	1622	0.310299	-0.0012002	-0.218524	0.261252
##	1631	0.005370	-0.1115340	-0.064250	-0.303676
##	1642	-0.293864	0.1227631	0.130031	0.132033
##	1652	-0.293864	0.1227631	0.130031	0.132033
##	1662	-0.025079	-0.0876930	-0.044481	-0.259341
##	1672	-0.293864	0.1227631	0.130031	0.132033
##	1682	0.851454	1.1015362	0.261615	-0.143930
##	1692	-0.178105	0.0321246	0.054873	-0.036522
##	1702	0.190710	-0.0821080	-0.162732	-0.069081
##	1714	0.087327	-0.0011601	-0.095609	0.081453
##	1722	0.390644	-0.8414047	1.200076	-0.023549
##	1731	0.005370	-0.1115340	-0.064250	-0.303676

## 1742	-0.293864	0.1227631	0.130031	0.132033
## 1752	0.005370	-0.1115340	-0.064250	-0.303676
## 1762	-0.178105	0.0321246	0.054873	-0.036522
## 1772	-0.062345	-0.0585139	-0.020285	-0.205078
## 1782	0.233555	0.0588900	-0.168697	0.372999
## 1792	-0.293864	0.1227631	0.130031	0.132033
## 1801	0.129298	-0.0340229	-0.122859	0.020340
## 1814	0.363980	-0.5438791	1.252024	0.814836
## 1822	-0.293864	0.1227631	0.130031	0.132033
## 1832	0.258269	-0.0329038	-0.193813	0.127641
## 1842	-0.293864	0.1227631	0.130031	0.132033
## 1852	-0.293864	0.1227631	0.130031	0.132033
## 1862	-0.110390	-0.0208955	0.010908	-0.135121
## 1872	-0.293864	0.1227631	0.130031	0.132033
## 1881	0.223592	-0.0057521	-0.171298	0.178133
## 1891	0.005370	-0.1115340	-0.064250	-0.303676
## 1901	-0.293864	0.1227631	0.130031	0.132033
## 1913	0.162811	-0.0602634	-0.144618	-0.028458
## 1921	0.005370	-0.1115340	-0.064250	-0.303676
## 1931	-0.178105	0.0321246	0.054873	-0.036522
## 1941	-0.293864	0.1227631	0.130031	0.132033
## 1951	-0.293864	0.1227631	0.130031	0.132033
## 1961	-0.025079	-0.0876930	-0.044481	-0.259341
## 1971	-0.293864	0.1227631	0.130031	0.132033
## 1981	0.005370	-0.1115340	-0.064250	-0.303676
## 1991	0.177342	-0.0716413	-0.154053	-0.049617
## 2001	-0.054177	0.1096357	-0.003736	0.287495
## 2013	-0.293864	0.1227631	0.130031	0.132033
## 2021	-0.025079	-0.0876930	-0.044481	-0.259341
## 2031	1.209922	2.3895810	1.009041	-1.063030
## 2041	0.223592	-0.0057521	-0.171298	0.178133
## 2051	1.182023	2.4114256	1.027155	-1.022407
## 2061	0.146893	-0.0478003	-0.134283	-0.005281
## 2071	0.109627	-0.0186212	-0.110088	0.048982
## 2081	-0.293864	0.1227631	0.130031	0.132033
## 2091	-0.110390	-0.0208955	0.010908	-0.135121
## 2103	0.031134	0.0428383	-0.059125	0.163275
## 2113	0.031114	-0.1316914	-0.080965	-0.341162
## 2121	-0.293864	0.1227631	0.130031	0.132033
## 2131	-0.025079	-0.0876930	-0.044481	-0.259341
## 2141	-0.062345	-0.0585139	-0.020285	-0.205078
## 2151	0.281599	0.0212716	-0.199890	0.303042
## 2161	-0.110390	-0.0208955	0.010908	-0.135121
## 2171	-0.293864	0.1227631	0.130031	0.132033
## 2181	-0.121892	0.1626558	0.040229	0.386093
## 2191	0.888585	-1.2051232	1.797326	0.542990
## 2201	0.005370	-0.1115340	-0.064250	-0.303676
## 2213	-0.293864	0.1227631	0.130031	0.132033
## 2221	0.314482	0.0976275	-0.208457	0.550256
## 2231	-0.293864	0.1227631	0.130031	0.132033
## 2241	-0.293864	0.1227631	0.130031	0.132033
## 2251	-0.054177	0.1096357	-0.003736	0.287495

##	2261	0.195693	0.0160925	-0.153184	0.218756
##	2271	-0.293864	0.1227631	0.130031	0.132033
##	2281	0.247491	-0.0244645	-0.186815	0.143335
##	2291	-0.293864	0.1227631	0.130031	0.132033
##	2301	0.087327	-0.0011601	-0.095609	0.081453
##	2313	0.005370	-0.1115340	-0.064250	-0.303676
##	2321	0.382197	0.0446073	-0.252422	0.451657
##	2331	0.247491	-0.0244645	-0.186815	0.143335
##	2341	0.005370	-0.1115340	-0.064250	-0.303676
##	2351	-0.293864	0.1227631	0.130031	0.132033
##	2361	0.342207	-0.0261833	-0.239241	0.214792
##	2371	-0.110390	-0.0208955	0.010908	-0.135121
##	2381	0.031114	-0.1316914	-0.080965	-0.341162
##	2391	-0.293864	0.1227631	0.130031	0.132033
##	2401	-0.293864	0.1227631	0.130031	0.132033
##	2413	-0.025079	-0.0876930	-0.044481	-0.259341
##	2421	-0.062345	-0.0585139	-0.020285	-0.205078
##	2431	0.162811	-0.0602634	-0.144618	-0.028458
##	2441	0.195693	0.0160925	-0.153184	0.218756
##	2451	-0.110390	-0.0208955	0.010908	-0.135121
##	2461	-0.110390	-0.0208955	0.010908	-0.135121
##	2471	0.031114	0.0428383	-0.059125	0.163275
##	2481	0.162180	0.0423330	-0.131426	0.267554
##	2491	0.005370	-0.1115340	-0.064250	-0.303676
##	2501	-0.293864	0.1227631	0.130031	0.132033
##	2513	-0.293864	0.1227631	0.130031	0.132033
##	2521	0.146893	-0.0478003	-0.134283	-0.005281
##	253	0.375058	-0.0519059	-0.260570	0.166958
##	254	-0.006133	0.0720173	-0.034930	0.217538
##	255	-0.062345	-0.0585139	-0.020285	-0.205078
##	256	0.342207	-0.0261833	-0.239241	0.214792
##	257	-0.025079	-0.0876930	-0.044481	-0.259341
##	258	0.061582	0.0189972	-0.078894	0.118939
##	259	0.087327	-0.0011601	-0.095609	0.081453
##	260	0.109627	-0.0186212	-0.110088	0.048982
##	261	0.031114	-0.1316914	-0.080965	-0.341162
##	262	0.342207	-0.0261833	-0.239241	0.214792
##	263	0.005370	-0.1115340	-0.064250	-0.303676
##	264	0.005370	-0.1115340	-0.064250	-0.303676
##	265	-0.025079	-0.0876930	-0.044481	-0.259341
##	266	0.309821	0.0553658	-0.211178	0.424351
##	267	0.061582	0.0189972	-0.078894	0.118939
##	268	0.005370	-0.1115340	-0.064250	-0.303676
##	269	0.268394	-0.0408313	-0.200386	0.112898
##	270	0.190710	-0.0821080	-0.162732	-0.069081
##	271	0.295535	-0.0620829	-0.218008	0.073378
##	272	-0.062345	-0.0585139	-0.020285	-0.205078
##	273	-0.293864	0.1227631	0.130031	0.132033
##	274	0.382197	0.0446073	-0.252422	0.451657
##	275	0.349314	-0.0317486	-0.243855	0.204443
##	276	-0.062345	-0.0585139	-0.020285	-0.205078
##	277	-0.293864	0.1227631	0.130031	0.132033

## 278	0.527784	-0.6721360	1.145672	0.576324
## 279	0.087327	-0.0011601	-0.095609	0.081453
## 280	-0.025079	-0.0876930	-0.044481	-0.259341
## 281	-0.293864	0.1227631	0.130031	0.132033
## 282	0.223592	-0.0057521	-0.171298	0.178133
## 283	0.177342	-0.0716413	-0.154053	-0.049617
## 284	0.177342	-0.0716413	-0.154053	-0.049617
## 285	0.005370	-0.1115340	-0.064250	-0.303676
## 286	0.005370	-0.1115340	-0.064250	-0.303676
## 287	-0.062345	-0.0585139	-0.020285	-0.205078
## 288	-0.062345	-0.0585139	-0.020285	-0.205078
## 289	-0.110390	-0.0208955	0.010908	-0.135121
## 290	0.061582	0.0189972	-0.078894	0.118939
## 291	0.031134	0.0428383	-0.059125	0.163275
## 292	0.005370	-0.1115340	-0.064250	-0.303676
## 293	0.005370	-0.1115340	-0.064250	-0.303676
## 294	-0.025079	-0.0876930	-0.044481	-0.259341
## 295	0.301270	0.0058698	-0.212662	0.274400
## 296	-0.293864	0.1227631	0.130031	0.132033
## 297	-0.025079	-0.0876930	-0.044481	-0.259341
## 298	-0.025079	-0.0876930	-0.044481	-0.259341
## 299	0.213884	0.0742917	-0.155926	0.401640
## 300	0.146893	-0.0478003	-0.134283	-0.005281
## 301	-0.110390	-0.0208955	0.010908	-0.135121
## 302	0.162811	-0.0602634	-0.144618	-0.028458
## 303	-0.293864	0.1227631	0.130031	0.132033
## 304	0.061582	0.0189972	-0.078894	0.118939
## 305	-0.293864	0.1227631	0.130031	0.132033
## 306	-0.062345	-0.0585139	-0.020285	-0.205078
## 307	0.146893	-0.0478003	-0.134283	-0.005281
## 308	-0.178105	0.0321246	0.054873	-0.036522
## 309	0.005370	-0.1115340	-0.064250	-0.303676
## 310	0.031134	0.0428383	-0.059125	0.163275
## 311	0.268394	-0.0408313	-0.200386	0.112898
## 312	-0.025079	-0.0876930	-0.044481	-0.259341
## 313	-0.025079	-0.0876930	-0.044481	-0.259341
## 314	-0.293864	0.1227631	0.130031	0.132033
## 315	0.235968	-0.0154427	-0.179334	0.160112
## 316	-0.025079	-0.0876930	-0.044481	-0.259341
## 317	-0.293864	0.1227631	0.130031	0.132033
## 318	-0.293864	0.1227631	0.130031	0.132033
## 319	0.061582	0.0189972	-0.078894	0.118939
## 320	-0.293864	0.1227631	0.130031	0.132033
## 321	0.005370	-0.1115340	-0.064250	-0.303676
## 322	-0.110390	-0.0208955	0.010908	-0.135121
## 323	0.177342	-0.0716413	-0.154053	-0.049617
## 324	-0.062345	-0.0585139	-0.020285	-0.205078
## 325	-0.178105	0.0321246	0.054873	-0.036522
## 326	0.109627	-0.0186212	-0.110088	0.048982
## 327	-0.062345	-0.0585139	-0.020285	-0.205078
## 328	0.005370	-0.1115340	-0.064250	-0.303676
## 329	-0.025079	-0.0876930	-0.044481	-0.259341

## 330	-0.293864	0.1227631	0.130031	0.132033
## 331	-0.025079	-0.0876930	-0.044481	-0.259341
## 332	-0.110390	-0.0208955	0.010908	-0.135121
## 333	0.031114	-0.1316914	-0.080965	-0.341162
## 334	0.031114	-0.1316914	-0.080965	-0.341162
## 335	0.475955	-0.9082022	1.144687	-0.147769
## 336	-0.293864	0.1227631	0.130031	0.132033
## 337	0.005370	-0.1115340	-0.064250	-0.303676
## 338	0.268394	-0.0408313	-0.200386	0.112898
## 339	-0.293864	0.1227631	0.130031	0.132033
## 340	-0.293864	0.1227631	0.130031	0.132033
## 341	0.395564	0.0341407	-0.261101	0.432193
## 342	0.005370	-0.1115340	-0.064250	-0.303676
## 343	0.005370	-0.1115340	-0.064250	-0.303676
## 344	-0.293864	0.1227631	0.130031	0.132033
## 345	-0.025079	-0.0876930	-0.044481	-0.259341
## 346	-0.025079	-0.0876930	-0.044481	-0.259341
## 347	0.203086	-0.0917986	-0.170767	-0.087102
## 348	0.295535	-0.0620829	-0.218008	0.073378
## 349	0.061582	0.0189972	-0.078894	0.118939
## 350	0.005370	-0.1115340	-0.064250	-0.303676
## 351	0.466989	0.0170336	-0.302614	0.440382
## 352	0.245057	-0.1246614	-0.198018	-0.148215
## 353	0.146893	-0.0478003	-0.134283	-0.005281
## 354	-0.178105	0.0321246	0.054873	-0.036522
## 355	0.179776	0.0285556	-0.142850	0.241933
## 356	0.210224	0.0047146	-0.162619	0.197597
## 357	0.005370	-0.1115340	-0.064250	-0.303676
## 358	-0.025079	-0.0876930	-0.044481	-0.259341
## 359	-0.293864	0.1227631	0.130031	0.132033
## 360	-0.025079	-0.0876930	-0.044481	-0.259341
## 361	-0.121892	0.1626558	0.040229	0.386093
## 362	-0.293864	0.1227631	0.130031	0.132033
## 363	0.270821	0.0297109	-0.192893	0.318736
## 364	-0.062345	-0.0585139	-0.020285	-0.205078
## 365	0.005370	-0.1115340	-0.064250	-0.303676
## 366	0.005370	-0.1115340	-0.064250	-0.303676
## 367	-0.293864	0.1227631	0.130031	0.132033
## 368	0.005370	-0.1115340	-0.064250	-0.303676
## 369	0.005370	-0.1115340	-0.064250	-0.303676
## 370	-0.110390	-0.0208955	0.010908	-0.135121
## 371	0.087327	-0.0011601	-0.095609	0.081453
## 372	-0.025079	-0.0876930	-0.044481	-0.259341
## 373	0.005370	-0.1115340	-0.064250	-0.303676
## 374	-0.110390	-0.0208955	0.010908	-0.135121
## 375	0.342207	-0.0261833	-0.239241	0.214792
## 376	0.005370	-0.1115340	-0.064250	-0.303676
## 377	-0.121892	0.1626558	0.040229	0.386093
## 378	-0.062345	-0.0585139	-0.020285	-0.205078
## 379	0.005370	-0.1115340	-0.064250	-0.303676
## 380	0.005370	-0.1115340	-0.064250	-0.303676
## 381	0.005370	-0.1115340	-0.064250	-0.303676

## 382	0.247491	-0.0244645	-0.186815	0.143335
## 383	0.005370	-0.1115340	-0.064250	-0.303676
## 384	0.268394	-0.0408313	-0.200386	0.112898
## 385	-0.293864	0.1227631	0.130031	0.132033
## 386	0.195693	0.0160925	-0.153184	0.218756
## 387	-0.293864	0.1227631	0.130031	0.132033
## 388	0.195693	0.0160925	-0.153184	0.218756
## 389	0.223592	-0.0057521	-0.171298	0.178133
## 390	-0.062345	-0.0585139	-0.020285	-0.205078
## 391	-0.025079	-0.0876930	-0.044481	-0.259341
## 392	-0.025079	-0.0876930	-0.044481	-0.259341
## 393	0.327014	-0.0142875	-0.229376	0.236914
## 394	0.268394	-0.0408313	-0.200386	0.112898
## 395	0.277939	-0.0483055	-0.206584	0.098999
## 396	-0.293864	0.1227631	0.130031	0.132033
## 397	-0.293864	0.1227631	0.130031	0.132033
## 398	-0.025079	-0.0876930	-0.044481	-0.259341
## 399	-0.025079	-0.0876930	-0.044481	-0.259341
## 400	-0.025079	-0.0876930	-0.044481	-0.259341
## 401	0.005370	-0.1115340	-0.064250	-0.303676
## 402	0.142509	0.0577347	-0.118654	0.296196
## 403	-0.293864	0.1227631	0.130031	0.132033
## 404	-0.025079	-0.0876930	-0.044481	-0.259341
## 405	-0.293864	0.1227631	0.130031	0.132033
## 406	0.177342	-0.0716413	-0.154053	-0.049617
## 407	0.146893	-0.0478003	-0.134283	-0.005281
## 408	0.268394	-0.0408313	-0.200386	0.112898
## 409	-0.293864	0.1227631	0.130031	0.132033
## 410	-0.062345	-0.0585139	-0.020285	-0.205078
## 411	-0.025079	-0.0876930	-0.044481	-0.259341
## 412	0.005370	-0.1115340	-0.064250	-0.303676
## 413	0.475955	-0.9082022	1.144687	-0.147769
## 414	0.177342	-0.0716413	-0.154053	-0.049617
## 415	0.005370	-0.1115340	-0.064250	-0.303676
## 416	-0.293864	0.1227631	0.130031	0.132033
## 417	0.162811	-0.0602634	-0.144618	-0.028458
## 418	-0.025079	-0.0876930	-0.044481	-0.259341
## 419	0.005370	-0.1115340	-0.064250	-0.303676
## 420	0.270821	0.0297109	-0.192893	0.318736
## 421	-0.293864	0.1227631	0.130031	0.132033
## 422	-0.293864	0.1227631	0.130031	0.132033
## 423	0.005370	-0.1115340	-0.064250	-0.303676
## 424	-0.025079	-0.0876930	-0.044481	-0.259341
## 425	-0.025079	-0.0876930	-0.044481	-0.259341
## 426	0.162811	-0.0602634	-0.144618	-0.028458
## 427	-0.110390	-0.0208955	0.010908	-0.135121
## 428	0.210224	0.0047146	-0.162619	0.197597
## 429	0.190710	-0.0821080	-0.162732	-0.069081
## 430	0.162811	-0.0602634	-0.144618	-0.028458
## 431	0.233555	0.0588900	-0.168697	0.372999
## 432	0.005370	-0.1115340	-0.064250	-0.303676
## 433	-0.062345	-0.0585139	-0.020285	-0.205078

## 434	0.177342	-0.0716413	-0.154053	-0.049617
## 435	0.177342	-0.0716413	-0.154053	-0.049617
## 436	-0.025079	-0.0876930	-0.044481	-0.259341
## 437	-0.025079	-0.0876930	-0.044481	-0.259341
## 438	0.393155	-0.0098842	-0.265284	0.303009
## 439	0.129298	-0.0340229	-0.122859	0.020340
## 440	-0.062345	-0.0585139	-0.020285	-0.205078
## 441	-0.293864	0.1227631	0.130031	0.132033
## 442	0.061582	0.0189972	-0.078894	0.118939
## 443	0.288918	0.0717325	-0.197607	0.454788
## 444	0.005370	-0.1115340	-0.064250	-0.303676
## 445	-0.110390	-0.0208955	0.010908	-0.135121
## 446	-0.062345	-0.0585139	-0.020285	-0.205078
## 447	0.177342	-0.0716413	-0.154053	-0.049617
## 448	-0.293864	0.1227631	0.130031	0.132033
## 449	-0.110390	-0.0208955	0.010908	-0.135121
## 450	0.247491	-0.0244645	-0.186815	0.143335
## 451	-0.062345	-0.0585139	-0.020285	-0.205078
## 452	0.129298	-0.0340229	-0.122859	0.020340
## 453	0.005370	-0.1115340	-0.064250	-0.303676
## 454	0.177342	-0.0716413	-0.154053	-0.049617
## 455	-0.110390	-0.0208955	0.010908	-0.135121
## 456	0.061582	0.0189972	-0.078894	0.118939
## 457	-0.293864	0.1227631	0.130031	0.132033
## 458	-0.293864	0.1227631	0.130031	0.132033
## 459	0.345110	0.0277342	-0.234091	0.372966
## 460	0.334152	0.0822257	-0.221228	0.521614
## 461	0.005370	-0.1115340	-0.064250	-0.303676
## 462	0.177342	-0.0716413	-0.154053	-0.049617
## 463	-0.110390	-0.0208955	0.010908	-0.135121
## 464	-0.025079	-0.0876930	-0.044481	-0.259341
## 465	0.005370	-0.1115340	-0.064250	-0.303676
## 466	0.393155	-0.0098842	-0.265284	0.303009
## 467	-0.062345	-0.0585139	-0.020285	-0.205078
## 468	-0.293864	0.1227631	0.130031	0.132033
## 469	0.005370	-0.1115340	-0.064250	-0.303676
## 470	0.223592	-0.0057521	-0.171298	0.178133
## 471	0.109627	-0.0186212	-0.110088	0.048982
## 472	0.061582	0.0189972	-0.078894	0.118939
## 473	-0.293864	0.1227631	0.130031	0.132033
## 474	-0.293864	0.1227631	0.130031	0.132033
## 475	-0.025079	-0.0876930	-0.044481	-0.259341
## 476	-0.178105	0.0321246	0.054873	-0.036522
## 477	-0.025079	-0.0876930	-0.044481	-0.259341
## 478	0.143539	0.1293711	-0.110253	0.504069
## 479	0.301270	0.0058698	-0.212662	0.274400
## 480	-0.293864	0.1227631	0.130031	0.132033
## 481	-0.178105	0.0321246	0.054873	-0.036522
## 482	0.087327	-0.0011601	-0.095609	0.081453
## 483	-0.293864	0.1227631	0.130031	0.132033
## 484	-0.293864	0.1227631	0.130031	0.132033
## 485	0.187924	0.0221757	-0.148140	0.230069

## 486	-0.293864	0.1227631	0.130031	0.132033
## 487	0.210224	0.0047146	-0.162619	0.197597
## 488	-0.062345	-0.0585139	-0.020285	-0.205078
## 489	0.005370	-0.1115340	-0.064250	-0.303676
## 490	-0.293864	0.1227631	0.130031	0.132033
## 491	0.839078	1.1112268	0.269651	-0.125909
## 492	-0.110390	-0.0208955	0.010908	-0.135121
## 493	0.129298	-0.0340229	-0.122859	0.020340
## 494	-0.293864	0.1227631	0.130031	0.132033
## 495	-0.025079	-0.0876930	-0.044481	-0.259341
## 496	-0.025079	-0.0876930	-0.044481	-0.259341
## 497	0.005370	-0.1115340	-0.064250	-0.303676
## 498	0.005370	-0.1115340	-0.064250	-0.303676
## 499	0.327014	-0.0142875	-0.229376	0.236914
## 500	-0.293864	0.1227631	0.130031	0.132033
## 501	0.195693	0.0160925	-0.153184	0.218756
## 502	-0.293864	0.1227631	0.130031	0.132033
## 503	0.223592	-0.0057521	-0.171298	0.178133
## 504	-0.062345	-0.0585139	-0.020285	-0.205078
## 505	0.005370	-0.1115340	-0.064250	-0.303676
## 506	-0.025079	-0.0876930	-0.044481	-0.259341
## 507	0.277939	-0.0483055	-0.206584	0.098999
## 508	-0.025079	-0.0876930	-0.044481	-0.259341
## 509	0.318866	-0.0079075	-0.224086	0.248779
## 510	-0.121892	0.1626558	0.040229	0.386093
## 511	-0.293864	0.1227631	0.130031	0.132033
## 512	-0.293864	0.1227631	0.130031	0.132033
## 513	-0.293864	0.1227631	0.130031	0.132033
## 514	0.120209	0.0751958	-0.104175	0.328668
## 515	-0.021295	0.1859916	-0.012302	0.534709
## 516	0.438689	-0.8790231	1.168883	-0.093506
## 517	0.416388	-0.8615620	1.183362	-0.061035
## 518	0.005370	-0.1115340	-0.064250	-0.303676
## 519	0.351748	0.0684484	-0.232652	0.495993
## 520	0.142509	0.0577347	-0.118654	0.296196
## 521	-0.293864	0.1227631	0.130031	0.132033
## 522	0.031114	-0.1316914	-0.080965	-0.341162
## 523	0.087327	-0.0011601	-0.095609	0.081453
## 524	0.005370	-0.1115340	-0.064250	-0.303676
## 525	-0.293864	0.1227631	0.130031	0.132033
## 526	-0.293864	0.1227631	0.130031	0.132033
## 527	-0.110390	-0.0208955	0.010908	-0.135121
## 528	-0.293864	0.1227631	0.130031	0.132033
## 529	0.061582	0.0189972	-0.078894	0.118939
## 530	0.005370	-0.1115340	-0.064250	-0.303676
## 531	-0.121892	0.1626558	0.040229	0.386093
## 532	0.318866	-0.0079075	-0.224086	0.248779
## 533	-0.293864	0.1227631	0.130031	0.132033
## 534	0.050080	0.2025486	-0.049574	0.640153
## 535	0.129298	-0.0340229	-0.122859	0.020340
## 536	-0.025079	-0.0876930	-0.044481	-0.259341
## 537	0.210224	0.0047146	-0.162619	0.197597

## 538	-0.293864	0.1227631	0.130031	0.132033
## 539	0.005370	-0.1115340	-0.064250	-0.303676
## 540	-0.006133	0.0720173	-0.034930	0.217538
## 541	0.309821	0.0553658	-0.211178	0.424351
## 542	0.005370	-0.1115340	-0.064250	-0.303676
## 543	-0.293864	0.1227631	0.130031	0.132033
## 544	-0.178105	0.0321246	0.054873	-0.036522
## 545	-0.293864	0.1227631	0.130031	0.132033
## 546	-0.025079	-0.0876930	-0.044481	-0.259341
## 547	0.031134	0.0428383	-0.059125	0.163275
## 548	-0.110390	-0.0208955	0.010908	-0.135121
## 549	0.162811	-0.0602634	-0.144618	-0.028458
## 550	-0.062345	-0.0585139	-0.020285	-0.205078
## 551	-0.025079	-0.0876930	-0.044481	-0.259341
## 552	0.247491	-0.0244645	-0.186815	0.143335
## 553	0.162180	0.0423330	-0.131426	0.267554
## 554	0.310299	-0.0012002	-0.218524	0.261252
## 555	-0.293864	0.1227631	0.130031	0.132033
## 556	-0.062345	-0.0585139	-0.020285	-0.205078
## 557	-0.025079	-0.0876930	-0.044481	-0.259341
## 558	0.246922	0.0484233	-0.177376	0.353534
## 559	-0.062345	-0.0585139	-0.020285	-0.205078
## 560	-0.025079	-0.0876930	-0.044481	-0.259341
## 561	0.334783	-0.0203706	-0.234421	0.225602
## 562	0.087327	-0.0011601	-0.095609	0.081453
## 563	0.005370	-0.1115340	-0.064250	-0.303676
## 564	0.177342	-0.0716413	-0.154053	-0.049617
## 565	0.162811	-0.0602634	-0.144618	-0.028458
## 566	0.910227	1.3285201	0.401018	-0.342567
## 567	-0.025079	-0.0876930	-0.044481	-0.259341
## 568	0.424939	0.0111405	-0.280173	0.389421
## 569	0.258269	-0.0329038	-0.193813	0.127641
## 570	0.005370	-0.1115340	-0.064250	-0.303676
## 571	0.177342	-0.0716413	-0.154053	-0.049617
## 572	0.258269	-0.0329038	-0.193813	0.127641
## 573	0.031114	-0.1316914	-0.080965	-0.341162
## 574	0.246922	0.0484233	-0.177376	0.353534
## 575	0.005370	-0.1115340	-0.064250	-0.303676
## 576	0.327014	-0.0142875	-0.229376	0.236914
## 577	-0.062345	-0.0585139	-0.020285	-0.205078
## 578	-0.006133	0.0720173	-0.034930	0.217538
## 579	0.005370	-0.1115340	-0.064250	-0.303676
## 580	0.005370	-0.1115340	-0.064250	-0.303676
## 581	-0.062345	-0.0585139	-0.020285	-0.205078
## 582	0.005370	-0.1115340	-0.064250	-0.303676
## 583	0.356132	-0.0370866	-0.248282	0.194516
## 584	0.301270	0.0058698	-0.212662	0.274400
## 585	-0.025079	-0.0876930	-0.044481	-0.259341
## 586	-0.054177	0.1096357	-0.003736	0.287495
## 587	0.258269	-0.0329038	-0.193813	0.127641
## 588	0.129298	-0.0340229	-0.122859	0.020340
## 589	-0.293864	0.1227631	0.130031	0.132033

## 590	0.475955	-0.9082022	1.144687	-0.147769
## 591	0.087327	-0.0011601	-0.095609	0.081453
## 592	-0.293864	0.1227631	0.130031	0.132033
## 593	0.005370	-0.1115340	-0.064250	-0.303676
## 594	-0.293864	0.1227631	0.130031	0.132033
## 595	-0.293864	0.1227631	0.130031	0.132033
## 596	-0.293864	0.1227631	0.130031	0.132033
## 597	0.179776	0.0285556	-0.142850	0.241933
## 598	0.087327	-0.0011601	-0.095609	0.081453
## 599	0.319366	0.0478915	-0.217376	0.410452
## 600	0.005370	-0.1115340	-0.064250	-0.303676
## 601	-0.293864	0.1227631	0.130031	0.132033
## 602	-0.293864	0.1227631	0.130031	0.132033
## 603	0.146893	-0.0478003	-0.134283	-0.005281
## 604	0.005370	-0.1115340	-0.064250	-0.303676
## 605	-0.293864	0.1227631	0.130031	0.132033
## 606	0.005370	-0.1115340	-0.064250	-0.303676
## 607	-0.293864	0.1227631	0.130031	0.132033
## 608	0.005370	-0.1115340	-0.064250	-0.303676
## 609	0.268394	-0.0408313	-0.200386	0.112898
## 610	0.258269	-0.0329038	-0.193813	0.127641
## 611	-0.293864	0.1227631	0.130031	0.132033
## 612	0.831734	1.3899795	0.451981	-0.228274
## 613	0.179776	0.0285556	-0.142850	0.241933
## 614	0.223592	-0.0057521	-0.171298	0.178133
## 615	-0.293864	0.1227631	0.130031	0.132033
## 616	0.005370	-0.1115340	-0.064250	-0.303676
## 617	-0.293864	0.1227631	0.130031	0.132033
## 618	-0.293864	0.1227631	0.130031	0.132033
## 619	0.031134	0.0428383	-0.059125	0.163275
## 620	-0.110390	-0.0208955	0.010908	-0.135121
## 621	0.162811	-0.0602634	-0.144618	-0.028458
## 622	-0.293864	0.1227631	0.130031	0.132033
## 623	-0.293864	0.1227631	0.130031	0.132033
## 624	-0.025079	-0.0876930	-0.044481	-0.259341
## 625	-0.025079	-0.0876930	-0.044481	-0.259341
## 626	0.532168	-0.7776709	1.130043	0.274847
## 627	0.177342	-0.0716413	-0.154053	-0.049617
## 628	0.190710	-0.0821080	-0.162732	-0.069081
## 629	0.281599	0.0212716	-0.199890	0.303042
## 630	0.258269	-0.0329038	-0.193813	0.127641
## 631	0.005370	-0.1115340	-0.064250	-0.303676
## 632	0.445207	-0.0047291	-0.293332	0.359909
## 633	-0.025079	-0.0876930	-0.044481	-0.259341
## 634	0.318866	-0.0079075	-0.224086	0.248779
## 635	-0.121892	0.1626558	0.040229	0.386093
## 636	-0.110390	-0.0208955	0.010908	-0.135121
## 637	0.318866	-0.0079075	-0.224086	0.248779
## 638	0.005370	-0.1115340	-0.064250	-0.303676
## 639	0.410153	-0.0231937	-0.276321	0.278258
## 640	0.031134	0.0428383	-0.059125	0.163275
## 641	0.318866	-0.0079075	-0.224086	0.248779

## 642	0.005370	-0.1115340	-0.064250	-0.303676
## 643	0.393155	-0.0098842	-0.265284	0.303009
## 644	0.435380	0.0029651	-0.286952	0.374217
## 645	0.087327	-0.0011601	-0.095609	0.081453
## 646	0.268394	-0.0408313	-0.200386	0.112898
## 647	0.031114	-0.1316914	-0.080965	-0.341162
## 648	0.031134	0.0428383	-0.059125	0.163275
## 649	-0.025079	-0.0876930	-0.044481	-0.259341
## 650	0.120209	0.0751958	-0.104175	0.328668
## 651	0.162811	-0.0602634	-0.144618	-0.028458
## 652	0.362682	-0.0422152	-0.252534	0.184979
## 653	0.005370	-0.1115340	-0.064250	-0.303676
## 654	0.094465	0.0953531	-0.087461	0.366153
## 655	0.109627	-0.0186212	-0.110088	0.048982
## 656	-0.293864	0.1227631	0.130031	0.132033
## 657	0.087327	-0.0011601	-0.095609	0.081453
## 658	0.031114	-0.1316914	-0.080965	-0.341162
## 659	0.223592	-0.0057521	-0.171298	0.178133
## 660	-0.121892	0.1626558	0.040229	0.386093
## 661	-0.025079	-0.0876930	-0.044481	-0.259341
## 662	0.247491	-0.0244645	-0.186815	0.143335
## 663	0.129298	-0.0340229	-0.122859	0.020340
## 664	0.177342	-0.0716413	-0.154053	-0.049617
## 665	0.223592	-0.0057521	-0.171298	0.178133
## 666	0.005370	-0.1115340	-0.064250	-0.303676
## 667	-0.025079	-0.0876930	-0.044481	-0.259341
## 668	0.087327	-0.0011601	-0.095609	0.081453
## 669	0.120209	0.0751958	-0.104175	0.328668
## 670	-0.025079	-0.0876930	-0.044481	-0.259341
## 671	0.310299	-0.0012002	-0.218524	0.261252
## 672	0.286969	-0.0553756	-0.212446	0.085851
## 673	-0.293864	0.1227631	0.130031	0.132033
## 674	0.466989	0.0170336	-0.302614	0.440382
## 675	0.005370	-0.1115340	-0.064250	-0.303676
## 676	0.179776	0.0285556	-0.142850	0.241933
## 677	-0.025079	-0.0876930	-0.044481	-0.259341
## 678	-0.025079	-0.0876930	-0.044481	-0.259341
## 679	-0.025079	-0.0876930	-0.044481	-0.259341
## 680	0.061582	0.0189972	-0.078894	0.118939
## 681	-0.110390	-0.0208955	0.010908	-0.135121
## 682	0.005370	-0.1115340	-0.064250	-0.303676
## 683	0.129298	-0.0340229	-0.122859	0.020340
## 684	0.778429	-0.7397549	0.999042	0.583137
## 685	0.380778	-0.0001936	-0.257249	0.321030
## 686	0.031114	-0.1316914	-0.080965	-0.341162
## 687	0.162811	-0.0602634	-0.144618	-0.028458
## 688	0.233555	0.0588900	-0.168697	0.372999
## 689	0.120209	0.0751958	-0.104175	0.328668
## 690	0.005370	-0.1115340	-0.064250	-0.303676
## 691	0.235968	-0.0154427	-0.179334	0.160112
## 692	-0.062345	-0.0585139	-0.020285	-0.205078
## 693	0.005370	-0.1115340	-0.064250	-0.303676

## 694	-0.025079	-0.0876930	-0.044481	-0.259341
## 695	0.680810	-0.7919536	1.046318	0.353505
## 696	-0.025079	-0.0876930	-0.044481	-0.259341
## 697	-0.293864	0.1227631	0.130031	0.132033
## 698	0.031114	-0.1316914	-0.080965	-0.341162
## 699	0.165840	0.1119101	-0.124732	0.471597
## 700	0.431964	0.0444574	-0.279874	0.491381
## 701	-0.025079	-0.0876930	-0.044481	-0.259341
## 702	0.387081	-0.0051286	-0.261341	0.311853
## 703	0.270821	0.0297109	-0.192893	0.318736
## 704	-0.293864	0.1227631	0.130031	0.132033
## 705	0.031134	0.0428383	-0.059125	0.163275
## 706	-0.025079	-0.0876930	-0.044481	-0.259341
## 707	-0.293864	0.1227631	0.130031	0.132033
## 708	0.061582	0.0189972	-0.078894	0.118939
## 709	-0.293864	0.1227631	0.130031	0.132033
## 710	-0.121892	0.1626558	0.040229	0.386093
## 711	0.005370	-0.1115340	-0.064250	-0.303676
## 712	0.129298	-0.0340229	-0.122859	0.020340
## 713	-0.293864	0.1227631	0.130031	0.132033
## 714	-0.062345	-0.0585139	-0.020285	-0.205078
## 715	-0.025079	-0.0876930	-0.044481	-0.259341
## 716	0.179776	0.0285556	-0.142850	0.241933
## 717	0.064016	0.1191941	-0.067691	0.410489
## 718	0.005370	-0.1115340	-0.064250	-0.303676
## 719	-0.062345	-0.0585139	-0.020285	-0.205078
## 720	0.749892	1.9020645	0.787890	-0.240466
## 721	0.281599	0.0212716	-0.199890	0.303042
## 722	0.005370	-0.1115340	-0.064250	-0.303676
## 723	-0.110390	-0.0208955	0.010908	-0.135121
## 724	0.457708	0.0243001	-0.296589	0.453895
## 725	-0.110390	-0.0208955	0.010908	-0.135121
## 726	-0.025079	-0.0876930	-0.044481	-0.259341
## 727	0.005370	-0.1115340	-0.064250	-0.303676
## 728	-0.293864	0.1227631	0.130031	0.132033
## 729	-0.293864	0.1227631	0.130031	0.132033
## 730	0.263382	-0.5672148	1.304555	0.666221
## 731	-0.293864	0.1227631	0.130031	0.132033
## 732	-0.062345	-0.0585139	-0.020285	-0.205078
## 733	0.345110	0.0277342	-0.234091	0.372966
## 734	0.005370	-0.1115340	-0.064250	-0.303676
## 735	-0.121892	0.1626558	0.040229	0.386093
## 736	-0.293864	0.1227631	0.130031	0.132033
## 737	-0.293864	0.1227631	0.130031	0.132033
## 738	-0.006133	0.0720173	-0.034930	0.217538
## 739	-0.062345	-0.0585139	-0.020285	-0.205078
## 740	-0.293864	0.1227631	0.130031	0.132033
## 741	-0.062345	-0.0585139	-0.020285	-0.205078
## 742	-0.293864	0.1227631	0.130031	0.132033
## 743	-0.293864	0.1227631	0.130031	0.132033
## 744	0.235968	-0.0154427	-0.179334	0.160112
## 745	-0.293864	0.1227631	0.130031	0.132033

## 746	0.270821	0.0297109	-0.192893	0.318736
## 747	-0.062345	-0.0585139	-0.020285	-0.205078
## 748	-0.293864	0.1227631	0.130031	0.132033
## 749	-0.293864	0.1227631	0.130031	0.132033
## 750	-0.025079	-0.0876930	-0.044481	-0.259341
## 751	0.005370	-0.1115340	-0.064250	-0.303676
## 752	-0.293864	0.1227631	0.130031	0.132033
## 753	-0.025079	-0.0876930	-0.044481	-0.259341
## 754	0.005370	-0.1115340	-0.064250	-0.303676
## 755	0.387081	-0.0051286	-0.261341	0.311853
## 756	0.005370	-0.1115340	-0.064250	-0.303676
## 757	0.868452	1.0882267	0.250579	-0.168681
## 758	-0.293864	0.1227631	0.130031	0.132033
## 759	-0.293864	0.1227631	0.130031	0.132033
## 760	-0.025079	-0.0876930	-0.044481	-0.259341
## 761	0.223592	-0.0057521	-0.171298	0.178133
## 762	1.106539	2.4705289	1.076164	-0.912496
## 763	0.177342	-0.0716413	-0.154053	-0.049617
## 764	0.129298	-0.0340229	-0.122859	0.020340
## 765	-0.293864	0.1227631	0.130031	0.132033
## 766	-0.121892	0.1626558	0.040229	0.386093
## 767	-0.062345	-0.0585139	-0.020285	-0.205078
## 768	-0.025079	-0.0876930	-0.044481	-0.259341
## 769	0.005370	-0.1115340	-0.064250	-0.303676
## 770	-0.293864	0.1227631	0.130031	0.132033
## 771	0.177342	-0.0716413	-0.154053	-0.049617
## 772	0.005370	-0.1115340	-0.064250	-0.303676
## 773	-0.054177	0.1096357	-0.003736	0.287495
## 774	0.258269	-0.0329038	-0.193813	0.127641
## 775	-0.293864	0.1227631	0.130031	0.132033
## 776	-0.025079	-0.0876930	-0.044481	-0.259341
## 777	-0.062345	-0.0585139	-0.020285	-0.205078
## 778	-0.062345	-0.0585139	-0.020285	-0.205078
## 779	-0.062345	-0.0585139	-0.020285	-0.205078
## 780	0.087327	-0.0011601	-0.095609	0.081453
## 781	0.005370	-0.1115340	-0.064250	-0.303676
## 782	0.087327	-0.0011601	-0.095609	0.081453
## 783	-0.293864	0.1227631	0.130031	0.132033
## 784	-0.293864	0.1227631	0.130031	0.132033
## 785	0.246922	0.0484233	-0.177376	0.353534
## 786	0.005370	-0.1115340	-0.064250	-0.303676
## 787	0.177342	-0.0716413	-0.154053	-0.049617
## 788	-0.293864	0.1227631	0.130031	0.132033
## 789	0.195693	0.0160925	-0.153184	0.218756
## 790	0.005370	-0.1115340	-0.064250	-0.303676
## 791	0.005370	-0.1115340	-0.064250	-0.303676
## 792	-0.293864	0.1227631	0.130031	0.132033
## 793	0.259299	0.0387326	-0.185412	0.335513
## 794	0.005370	-0.1115340	-0.064250	-0.303676
## 795	-0.110390	-0.0208955	0.010908	-0.135121
## 796	0.399015	-0.0144729	-0.269089	0.294476
## 797	-0.293864	0.1227631	0.130031	0.132033

## 798	-0.293864	0.1227631	0.130031	0.132033
## 799	-0.062345	-0.0585139	-0.020285	-0.205078
## 800	0.005370	-0.1115340	-0.064250	-0.303676
## 801	0.268394	-0.0408313	-0.200386	0.112898
## 802	0.223592	-0.0057521	-0.171298	0.178133
## 803	-0.025079	-0.0876930	-0.044481	-0.259341
## 804	-0.025079	-0.0876930	-0.044481	-0.259341
## 805	-0.110390	-0.0208955	0.010908	-0.135121
## 806	-0.293864	0.1227631	0.130031	0.132033
## 807	-0.025079	-0.0876930	-0.044481	-0.259341
## 808	0.005370	-0.1115340	-0.064250	-0.303676
## 809	-0.110390	-0.0208955	0.010908	-0.135121
## 810	-0.110390	-0.0208955	0.010908	-0.135121
## 811	0.005370	-0.1115340	-0.064250	-0.303676
## 812	0.031114	-0.1316914	-0.080965	-0.341162
## 813	0.129298	-0.0340229	-0.122859	0.020340
## 814	0.146893	-0.0478003	-0.134283	-0.005281
## 815	-0.293864	0.1227631	0.130031	0.132033
## 816	0.233555	0.0588900	-0.168697	0.372999
## 817	0.223592	-0.0057521	-0.171298	0.178133
## 818	-0.293864	0.1227631	0.130031	0.132033
## 819	0.005370	-0.1115340	-0.064250	-0.303676
## 820	0.005370	-0.1115340	-0.064250	-0.303676
## 821	-0.293864	0.1227631	0.130031	0.132033
## 822	-0.025079	-0.0876930	-0.044481	-0.259341
## 823	-0.178105	0.0321246	0.054873	-0.036522
## 824	-0.293864	0.1227631	0.130031	0.132033
## 825	-0.062345	-0.0585139	-0.020285	-0.205078
## 826	0.005370	-0.1115340	-0.064250	-0.303676
## 827	0.146893	-0.0478003	-0.134283	-0.005281
## 828	-0.293864	0.1227631	0.130031	0.132033
## 829	-0.062345	-0.0585139	-0.020285	-0.205078
## 830	-0.025079	-0.0876930	-0.044481	-0.259341
## 831	-0.025079	-0.0876930	-0.044481	-0.259341
## 832	-0.121892	0.1626558	0.040229	0.386093
## 833	-0.293864	0.1227631	0.130031	0.132033
## 834	-0.121892	0.1626558	0.040229	0.386093
## 835	-0.178105	0.0321246	0.054873	-0.036522
## 836	-0.025079	-0.0876930	-0.044481	-0.259341
## 837	-0.025079	-0.0876930	-0.044481	-0.259341
## 838	-0.293864	0.1227631	0.130031	0.132033
## 839	-0.062345	-0.0585139	-0.020285	-0.205078
## 840	-0.293864	0.1227631	0.130031	0.132033
## 841	0.146893	-0.0478003	-0.134283	-0.005281
## 842	-0.121892	0.1626558	0.040229	0.386093
## 843	0.129298	-0.0340229	-0.122859	0.020340
## 844	0.005370	-0.1115340	-0.064250	-0.303676
## 845	0.322929	-0.7883846	1.244041	0.075050
## 846	0.491872	-0.9206653	1.134353	-0.170946
## 847	-0.006133	0.0720173	-0.034930	0.217538
## 848	-0.062345	-0.0585139	-0.020285	-0.205078
## 849	-0.025079	-0.0876930	-0.044481	-0.259341

## 850	0.005370	-0.1115340	-0.064250	-0.303676
## 851	-0.025079	-0.0876930	-0.044481	-0.259341
## 852	0.129298	-0.0340229	-0.122859	0.020340
## 853	0.374228	0.0049351	-0.252996	0.330568
## 854	-0.110390	-0.0208955	0.010908	-0.135121
## 855	0.005370	-0.1115340	-0.064250	-0.303676
## 856	-0.025079	-0.0876930	-0.044481	-0.259341
## 857	-0.121892	0.1626558	0.040229	0.386093
## 858	-0.025079	-0.0876930	-0.044481	-0.259341
## 859	0.233555	0.0588900	-0.168697	0.372999
## 860	0.109627	-0.0186212	-0.110088	0.048982
## 861	0.087327	-0.0011601	-0.095609	0.081453
## 862	0.087327	-0.0011601	-0.095609	0.081453
## 863	-0.293864	0.1227631	0.130031	0.132033
## 864	0.005370	-0.1115340	-0.064250	-0.303676
## 865	-0.293864	0.1227631	0.130031	0.132033
## 866	-0.293864	0.1227631	0.130031	0.132033
## 867	-0.293864	0.1227631	0.130031	0.132033
## 868	-0.025079	-0.0876930	-0.044481	-0.259341
## 869	-0.293864	0.1227631	0.130031	0.132033
## 870	-0.293864	0.1227631	0.130031	0.132033
## 871	-0.178105	0.0321246	0.054873	-0.036522
## 872	0.005370	-0.1115340	-0.064250	-0.303676
## 873	-0.062345	-0.0585139	-0.020285	-0.205078
## 874	0.129298	-0.0340229	-0.122859	0.020340
## 875	0.005370	-0.1115340	-0.064250	-0.303676
## 876	-0.293864	0.1227631	0.130031	0.132033
## 877	-0.178105	0.0321246	0.054873	-0.036522
## 878	-0.121892	0.1626558	0.040229	0.386093
## 879	0.195693	0.0160925	-0.153184	0.218756
## 880	-0.062345	-0.0585139	-0.020285	-0.205078
## 881	-0.293864	0.1227631	0.130031	0.132033
## 882	0.005370	-0.1115340	-0.064250	-0.303676
## 883	0.258269	-0.0329038	-0.193813	0.127641
## 884	0.005370	-0.1115340	-0.064250	-0.303676
## 885	-0.293864	0.1227631	0.130031	0.132033
## 886	-0.293864	0.1227631	0.130031	0.132033
## 887	-0.293864	0.1227631	0.130031	0.132033
## 888	-0.121892	0.1626558	0.040229	0.386093
## 889	0.342207	-0.0261833	-0.239241	0.214792
## 890	0.031134	0.0428383	-0.059125	0.163275
## 891	-0.293864	0.1227631	0.130031	0.132033
## 892	0.005370	-0.1115340	-0.064250	-0.303676
## 893	-0.178105	0.0321246	0.054873	-0.036522
## 894	0.005370	-0.1115340	-0.064250	-0.303676
## 895	0.087327	-0.0011601	-0.095609	0.081453
## 896	0.268394	-0.0408313	-0.200386	0.112898
## 897	-0.293864	0.1227631	0.130031	0.132033
## 898	0.087327	-0.0011601	-0.095609	0.081453
## 899	-0.025079	-0.0876930	-0.044481	-0.259341
## 900	-0.293864	0.1227631	0.130031	0.132033
## 901	-0.293864	0.1227631	0.130031	0.132033

## 902	0.162811	-0.0602634	-0.144618	-0.028458
## 903	0.177342	-0.0716413	-0.154053	-0.049617
## 904	-0.293864	0.1227631	0.130031	0.132033
## 905	0.610661	-0.8391304	1.079080	0.160554
## 906	0.031114	-0.1316914	-0.080965	-0.341162
## 907	0.223592	-0.0057521	-0.171298	0.178133
## 908	0.087327	-0.0011601	-0.095609	0.081453
## 909	0.005370	-0.1115340	-0.064250	-0.303676
## 910	0.005370	-0.1115340	-0.064250	-0.303676
## 911	-0.121892	0.1626558	0.040229	0.386093
## 912	-0.006133	0.0720173	-0.034930	0.217538
## 913	-0.293864	0.1227631	0.130031	0.132033
## 914	0.026750	0.1483732	-0.043496	0.464752
## 915	-0.293864	0.1227631	0.130031	0.132033
## 916	-0.121892	0.1626558	0.040229	0.386093
## 917	-0.293864	0.1227631	0.130031	0.132033
## 918	-0.293864	0.1227631	0.130031	0.132033
## 919	-0.178105	0.0321246	0.054873	-0.036522
## 920	0.142509	0.0577347	-0.118654	0.296196
## 921	0.064016	0.1191941	-0.067691	0.410489
## 922	0.005370	-0.1115340	-0.064250	-0.303676
## 923	-0.025079	-0.0876930	-0.044481	-0.259341
## 924	-0.062345	-0.0585139	-0.020285	-0.205078
## 925	-0.293864	0.1227631	0.130031	0.132033
## 926	0.026750	0.1483732	-0.043496	0.464752
## 927	-0.293864	0.1227631	0.130031	0.132033
## 928	0.120209	0.0751958	-0.104175	0.328668
## 929	0.087327	-0.0011601	-0.095609	0.081453
## 930	-0.293864	0.1227631	0.130031	0.132033
## 931	-0.025079	-0.0876930	-0.044481	-0.259341
## 932	-0.293864	0.1227631	0.130031	0.132033
## 933	0.094465	0.0953531	-0.087461	0.366153
## 934	-0.293864	0.1227631	0.130031	0.132033
## 935	0.179776	0.0285556	-0.142850	0.241933
## 936	-0.178105	0.0321246	0.054873	-0.036522
## 937	0.005370	-0.1115340	-0.064250	-0.303676
## 938	0.129298	-0.0340229	-0.122859	0.020340
## 939	-0.121892	0.1626558	0.040229	0.386093
## 940	0.146893	-0.0478003	-0.134283	-0.005281
## 941	-0.121892	0.1626558	0.040229	0.386093
## 942	0.146893	-0.0478003	-0.134283	-0.005281
## 943	-0.293864	0.1227631	0.130031	0.132033
## 944	0.210224	0.0047146	-0.162619	0.197597
## 945	0.162180	0.0423330	-0.131426	0.267554
## 946	-0.293864	0.1227631	0.130031	0.132033
## 947	-0.293864	0.1227631	0.130031	0.132033
## 948	-0.293864	0.1227631	0.130031	0.132033
## 949	-0.293864	0.1227631	0.130031	0.132033
## 950	-0.025079	-0.0876930	-0.044481	-0.259341
## 951	-0.293864	0.1227631	0.130031	0.132033
## 952	-0.293864	0.1227631	0.130031	0.132033
## 953	-0.025079	-0.0876930	-0.044481	-0.259341

## 954	-0.025079	-0.0876930	-0.044481	-0.259341
## 955	-0.062345	-0.0585139	-0.020285	-0.205078
## 956	-0.025079	-0.0876930	-0.044481	-0.259341
## 957	-0.062345	-0.0585139	-0.020285	-0.205078
## 958	-0.293864	0.1227631	0.130031	0.132033
## 959	-0.025079	-0.0876930	-0.044481	-0.259341
## 960	-0.110390	-0.0208955	0.010908	-0.135121
## 961	0.146893	-0.0478003	-0.134283	-0.005281
## 962	-0.025079	-0.0876930	-0.044481	-0.259341
## 963	0.031134	0.0428383	-0.059125	0.163275
## 964	-0.293864	0.1227631	0.130031	0.132033
## 965	-0.021295	0.1859916	-0.012302	0.534709
## 966	-0.121892	0.1626558	0.040229	0.386093
## 967	-0.062345	-0.0585139	-0.020285	-0.205078
## 968	-0.121892	0.1626558	0.040229	0.386093
## 969	-0.293864	0.1227631	0.130031	0.132033
## 970	-0.293864	0.1227631	0.130031	0.132033
## 971	0.177342	-0.0716413	-0.154053	-0.049617
## 972	-0.062345	-0.0585139	-0.020285	-0.205078
## 973	-0.110390	-0.0208955	0.010908	-0.135121
## 974	-0.293864	0.1227631	0.130031	0.132033
## 975	-0.110390	-0.0208955	0.010908	-0.135121
## 976	0.087327	-0.0011601	-0.095609	0.081453
## 977	-0.062345	-0.0585139	-0.020285	-0.205078
## 978	0.005370	-0.1115340	-0.064250	-0.303676
## 979	0.031134	0.0428383	-0.059125	0.163275
## 980	-0.293864	0.1227631	0.130031	0.132033
## 981	0.235968	-0.0154427	-0.179334	0.160112
## 982	-0.293864	0.1227631	0.130031	0.132033
## 983	0.005370	-0.1115340	-0.064250	-0.303676
## 984	-0.293864	0.1227631	0.130031	0.132033
## 985	-0.293864	0.1227631	0.130031	0.132033
## 986	-0.025079	-0.0876930	-0.044481	-0.259341
## 987	-0.293864	0.1227631	0.130031	0.132033
## 988	-0.293864	0.1227631	0.130031	0.132033
## 989	-0.293864	0.1227631	0.130031	0.132033
## 990	-0.293864	0.1227631	0.130031	0.132033
## 991	0.146893	-0.0478003	-0.134283	-0.005281
## 992	-0.025079	-0.0876930	-0.044481	-0.259341
## 993	-0.293864	0.1227631	0.130031	0.132033
## 994	0.005370	-0.1115340	-0.064250	-0.303676
## 995	-0.025079	-0.0876930	-0.044481	-0.259341
## 996	-0.110390	-0.0208955	0.010908	-0.135121
## 997	-0.025079	-0.0876930	-0.044481	-0.259341
## 998	0.005370	-0.1115340	-0.064250	-0.303676
## 999	-0.293864	0.1227631	0.130031	0.132033
## 1000	-0.054177	0.1096357	-0.003736	0.287495
## 1001	0.142509	0.0577347	-0.118654	0.296196
## 1002	0.223592	-0.0057521	-0.171298	0.178133
## 1003	-0.062345	-0.0585139	-0.020285	-0.205078
## 1004	-0.293864	0.1227631	0.130031	0.132033
## 1005	0.129298	-0.0340229	-0.122859	0.020340

## 1006	0.109627	-0.0186212	-0.110088	0.048982
## 1007	-0.062345	-0.0585139	-0.020285	-0.205078
## 1008	0.419463	0.0154283	-0.276617	0.397394
## 1009	-0.293864	0.1227631	0.130031	0.132033
## 1010	-0.121892	0.1626558	0.040229	0.386093
## 1011	-0.293864	0.1227631	0.130031	0.132033
## 1012	-0.293864	0.1227631	0.130031	0.132033
## 1013	0.005370	-0.1115340	-0.064250	-0.303676
## 1014	0.162811	-0.0602634	-0.144618	-0.028458
## 1015	0.005370	-0.1115340	-0.064250	-0.303676
## 1016	-0.006133	0.0720173	-0.034930	0.217538
## 1017	-0.025079	-0.0876930	-0.044481	-0.259341
## 1018	-0.025079	-0.0876930	-0.044481	-0.259341
## 1019	-0.110390	-0.0208955	0.010908	-0.135121
## 1020	-0.025079	-0.0876930	-0.044481	-0.259341
## 1021	-0.062345	-0.0585139	-0.020285	-0.205078
## 1022	-0.025079	-0.0876930	-0.044481	-0.259341
## 1023	-0.293864	0.1227631	0.130031	0.132033
## 1024	1.222298	2.3798904	1.001006	-1.081051
## 1025	0.005370	-0.1115340	-0.064250	-0.303676
## 1026	-0.025079	-0.0876930	-0.044481	-0.259341
## 1027	0.292181	0.1150885	-0.193978	0.582727
## 1028	0.061582	0.0189972	-0.078894	0.118939
## 1029	0.162811	-0.0602634	-0.144618	-0.028458
## 1030	-0.062345	-0.0585139	-0.020285	-0.205078
## 1031	0.268394	-0.0408313	-0.200386	0.112898
## 1032	0.251151	0.0451126	-0.180121	0.347378
## 1033	-0.025079	-0.0876930	-0.044481	-0.259341
## 1034	-0.293864	0.1227631	0.130031	0.132033
## 1035	-0.025079	-0.0876930	-0.044481	-0.259341
## 1036	0.005370	-0.1115340	-0.064250	-0.303676
## 1037	0.310299	-0.0012002	-0.218524	0.261252
## 1038	0.162180	0.0423330	-0.131426	0.267554
## 1039	-0.293864	0.1227631	0.130031	0.132033
## 1040	-0.293864	0.1227631	0.130031	0.132033
## 1041	-0.110390	-0.0208955	0.010908	-0.135121
## 1042	0.129298	-0.0340229	-0.122859	0.020340
## 1043	0.268394	-0.0408313	-0.200386	0.112898
## 1044	-0.110390	-0.0208955	0.010908	-0.135121
## 1045	0.286969	-0.0553756	-0.212446	0.085851
## 1046	0.109627	-0.0186212	-0.110088	0.048982
## 1047	0.195693	0.0160925	-0.153184	0.218756
## 1048	0.309821	0.0553658	-0.211178	0.424351
## 1049	0.031114	-0.1316914	-0.080965	-0.341162
## 1050	0.005370	-0.1115340	-0.064250	-0.303676
## 1051	0.277939	-0.0483055	-0.206584	0.098999
## 1052	0.548085	-0.7901340	1.119708	0.251670
## 1053	0.005370	-0.1115340	-0.064250	-0.303676
## 1054	0.005370	-0.1115340	-0.064250	-0.303676
## 1055	-0.025079	-0.0876930	-0.044481	-0.259341
## 1056	-0.025079	-0.0876930	-0.044481	-0.259341
## 1057	0.061582	0.0189972	-0.078894	0.118939

## 1058	-0.293864	0.1227631	0.130031	0.132033
## 1059	-0.293864	0.1227631	0.130031	0.132033
## 1060	-0.293864	0.1227631	0.130031	0.132033
## 1061	0.005370	-0.1115340	-0.064250	-0.303676
## 1062	0.061582	0.0189972	-0.078894	0.118939
## 1063	0.210224	0.0047146	-0.162619	0.197597
## 1064	-0.293864	0.1227631	0.130031	0.132033
## 1065	-0.062345	-0.0585139	-0.020285	-0.205078
## 1066	-0.025079	-0.0876930	-0.044481	-0.259341
## 1067	0.005370	-0.1115340	-0.064250	-0.303676
## 1068	0.162811	-0.0602634	-0.144618	-0.028458
## 1069	0.005370	-0.1115340	-0.064250	-0.303676
## 1070	0.087327	-0.0011601	-0.095609	0.081453
## 1071	-0.293864	0.1227631	0.130031	0.132033
## 1072	-0.293864	0.1227631	0.130031	0.132033
## 1073	-0.293864	0.1227631	0.130031	0.132033
## 1074	0.091410	-0.6071076	1.394358	0.412161
## 1075	-0.293864	0.1227631	0.130031	0.132033
## 1076	0.247491	-0.0244645	-0.186815	0.143335
## 1077	0.087327	-0.0011601	-0.095609	0.081453
## 1078	-0.293864	0.1227631	0.130031	0.132033
## 1079	-0.025079	-0.0876930	-0.044481	-0.259341
## 1080	-0.006133	0.0720173	-0.034930	0.217538
## 1081	-0.062345	-0.0585139	-0.020285	-0.205078
## 1082	0.438689	-0.8790231	1.168883	-0.093506
## 1083	-0.110390	-0.0208955	0.010908	-0.135121
## 1084	0.162811	-0.0602634	-0.144618	-0.028458
## 1085	-0.293864	0.1227631	0.130031	0.132033
## 1086	-0.293864	0.1227631	0.130031	0.132033
## 1087	0.005370	-0.1115340	-0.064250	-0.303676
## 1088	0.005370	-0.1115340	-0.064250	-0.303676
## 1089	-0.062345	-0.0585139	-0.020285	-0.205078
## 1090	0.005370	-0.1115340	-0.064250	-0.303676
## 1091	0.005370	-0.1115340	-0.064250	-0.303676
## 1092	-0.293864	0.1227631	0.130031	0.132033
## 1093	-0.025079	-0.0876930	-0.044481	-0.259341
## 1094	0.362682	-0.0422152	-0.252534	0.184979
## 1095	0.162811	-0.0602634	-0.144618	-0.028458
## 1096	-0.293864	0.1227631	0.130031	0.132033
## 1097	0.162811	-0.0602634	-0.144618	-0.028458
## 1098	0.177342	-0.0716413	-0.154053	-0.049617
## 1099	0.005370	-0.1115340	-0.064250	-0.303676
## 1100	-0.293864	0.1227631	0.130031	0.132033
## 1101	0.031114	-0.1316914	-0.080965	-0.341162
## 1102	0.129298	-0.0340229	-0.122859	0.020340
## 1103	-0.062345	-0.0585139	-0.020285	-0.205078
## 1104	-0.178105	0.0321246	0.054873	-0.036522
## 1105	-0.062345	-0.0585139	-0.020285	-0.205078
## 1106	-0.293864	0.1227631	0.130031	0.132033
## 1107	0.005370	-0.1115340	-0.064250	-0.303676
## 1108	-0.110390	-0.0208955	0.010908	-0.135121
## 1109	-0.293864	0.1227631	0.130031	0.132033

## 1110	0.352879	0.0216511	-0.239135	0.361654
## 1111	0.190710	-0.0821080	-0.162732	-0.069081
## 1112	-0.062345	-0.0585139	-0.020285	-0.205078
## 1113	-0.293864	0.1227631	0.130031	0.132033
## 1114	0.005370	-0.1115340	-0.064250	-0.303676
## 1115	-0.025079	-0.0876930	-0.044481	-0.259341
## 1116	-0.293864	0.1227631	0.130031	0.132033
## 1117	-0.110390	-0.0208955	0.010908	-0.135121
## 1118	-0.293864	0.1227631	0.130031	0.132033
## 1119	-0.025079	-0.0876930	-0.044481	-0.259341
## 1120	-0.062345	-0.0585139	-0.020285	-0.205078
## 1121	-0.025079	-0.0876930	-0.044481	-0.259341
## 1122	-0.062345	-0.0585139	-0.020285	-0.205078
## 1123	-0.293864	0.1227631	0.130031	0.132033
## 1124	-0.025079	-0.0876930	-0.044481	-0.259341
## 1125	-0.293864	0.1227631	0.130031	0.132033
## 1126	0.374228	0.0049351	-0.252996	0.330568
## 1127	-0.121892	0.1626558	0.040229	0.386093
## 1128	0.087327	-0.0011601	-0.095609	0.081453
## 1129	0.005370	-0.1115340	-0.064250	-0.303676
## 1130	-0.025079	-0.0876930	-0.044481	-0.259341
## 1131	-0.293864	0.1227631	0.130031	0.132033
## 1132	0.334783	-0.0203706	-0.234421	0.225602
## 1133	0.005370	-0.1115340	-0.064250	-0.303676
## 1134	0.334783	-0.0203706	-0.234421	0.225602
## 1135	0.005370	-0.1115340	-0.064250	-0.303676
## 1136	0.005370	-0.1115340	-0.064250	-0.303676
## 1137	0.005370	-0.1115340	-0.064250	-0.303676
## 1138	-0.293864	0.1227631	0.130031	0.132033
## 1139	-0.293864	0.1227631	0.130031	0.132033
## 1140	-0.025079	-0.0876930	-0.044481	-0.259341
## 1141	-0.293864	0.1227631	0.130031	0.132033
## 1142	-0.178105	0.0321246	0.054873	-0.036522
## 1143	-0.293864	0.1227631	0.130031	0.132033
## 1144	-0.293864	0.1227631	0.130031	0.132033
## 1145	-0.293864	0.1227631	0.130031	0.132033
## 1146	-0.293864	0.1227631	0.130031	0.132033
## 1147	0.005370	-0.1115340	-0.064250	-0.303676
## 1148	-0.293864	0.1227631	0.130031	0.132033
## 1149	0.339769	-0.0242743	-0.237658	0.218342
## 1150	-0.293864	0.1227631	0.130031	0.132033
## 1151	-0.293864	0.1227631	0.130031	0.132033
## 1152	0.268394	-0.0408313	-0.200386	0.112898
## 1153	0.162811	-0.0602634	-0.144618	-0.028458
## 1154	-0.025079	-0.0876930	-0.044481	-0.259341
## 1155	0.235968	-0.0154427	-0.179334	0.160112
## 1156	-0.293864	0.1227631	0.130031	0.132033
## 1157	0.277939	-0.0483055	-0.206584	0.098999
## 1158	-0.293864	0.1227631	0.130031	0.132033
## 1159	0.005370	-0.1115340	-0.064250	-0.303676
## 1160	-0.293864	0.1227631	0.130031	0.132033
## 1161	-0.293864	0.1227631	0.130031	0.132033

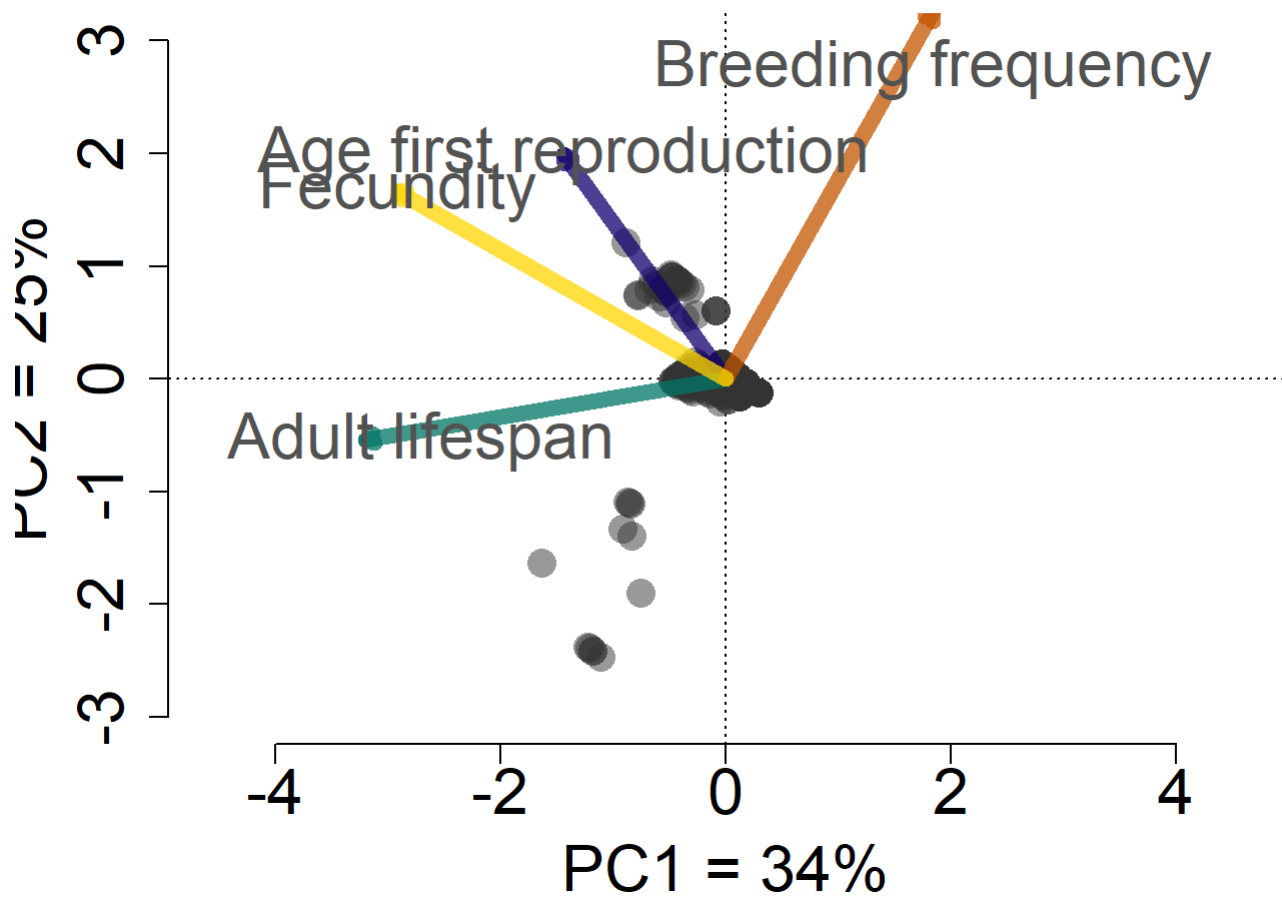
##	1162	-0.293864	0.1227631	0.130031	0.132033
##	1163	0.005370	-0.1115340	-0.064250	-0.303676
##	1164	-0.293864	0.1227631	0.130031	0.132033
##	1165	-0.121892	0.1626558	0.040229	0.386093
##	1166	-0.293864	0.1227631	0.130031	0.132033
##	1167	0.162811	-0.0602634	-0.144618	-0.028458
##	1168	-0.025079	-0.0876930	-0.044481	-0.259341
##	1169	0.146893	-0.0478003	-0.134283	-0.005281
##	1170	0.005370	-0.1115340	-0.064250	-0.303676
##	1171	-0.025079	-0.0876930	-0.044481	-0.259341
##	1172	-0.062345	-0.0585139	-0.020285	-0.205078
##	1173	-0.062345	-0.0585139	-0.020285	-0.205078
##	1174	-0.025079	-0.0876930	-0.044481	-0.259341
##	1175	0.005370	-0.1115340	-0.064250	-0.303676
##	1176	-0.293864	0.1227631	0.130031	0.132033
##	1177	-0.006133	0.0720173	-0.034930	0.217538
##	1178	0.177342	-0.0716413	-0.154053	-0.049617
##	1179	-0.025079	-0.0876930	-0.044481	-0.259341
##	1180	0.005370	-0.1115340	-0.064250	-0.303676
##	1181	-0.293864	0.1227631	0.130031	0.132033
##	1182	0.061582	0.0189972	-0.078894	0.118939
##	1183	-0.293864	0.1227631	0.130031	0.132033
##	1184	0.005370	-0.1115340	-0.064250	-0.303676
##	1185	0.146893	-0.0478003	-0.134283	-0.005281
##	1186	-0.293864	0.1227631	0.130031	0.132033
##	1187	-0.293864	0.1227631	0.130031	0.132033
##	1188	-0.293864	0.1227631	0.130031	0.132033
##	1189	0.162811	-0.0602634	-0.144618	-0.028458
##	1190	0.091410	-0.6071076	1.394358	0.412161
##	1191	-0.293864	0.1227631	0.130031	0.132033
##	1192	-0.062345	-0.0585139	-0.020285	-0.205078
##	1193	-0.110390	-0.0208955	0.010908	-0.135121
##	1194	-0.293864	0.1227631	0.130031	0.132033
##	1195	-0.110390	-0.0208955	0.010908	-0.135121
##	1196	-0.293864	0.1227631	0.130031	0.132033
##	1197	-0.293864	0.1227631	0.130031	0.132033
##	1198	-0.293864	0.1227631	0.130031	0.132033
##	1199	0.162180	0.0423330	-0.131426	0.267554
##	1200	-0.293864	0.1227631	0.130031	0.132033
##	1201	-0.293864	0.1227631	0.130031	0.132033
##	1202	-0.293864	0.1227631	0.130031	0.132033
##	1203	-0.054177	0.1096357	-0.003736	0.287495
##	1204	-0.293864	0.1227631	0.130031	0.132033
##	1205	-0.178105	0.0321246	0.054873	-0.036522
##	1206	-0.110390	-0.0208955	0.010908	-0.135121
##	1207	-0.062345	-0.0585139	-0.020285	-0.205078
##	1208	-0.025079	-0.0876930	-0.044481	-0.259341
##	1209	0.005370	-0.1115340	-0.064250	-0.303676
##	1210	-0.025079	-0.0876930	-0.044481	-0.259341
##	1211	-0.293864	0.1227631	0.130031	0.132033
##	1212	-0.293864	0.1227631	0.130031	0.132033
##	1213	-0.293864	0.1227631	0.130031	0.132033

##	1214	-0.293864	0.1227631	0.130031	0.132033
##	1215	-0.006133	0.0720173	-0.034930	0.217538
##	1216	-0.025079	-0.0876930	-0.044481	-0.259341
##	1217	-0.293864	0.1227631	0.130031	0.132033
##	1218	-0.293864	0.1227631	0.130031	0.132033
##	1219	-0.293864	0.1227631	0.130031	0.132033
##	1220	-0.293864	0.1227631	0.130031	0.132033
##	1221	-0.293864	0.1227631	0.130031	0.132033
##	1222	-0.062345	-0.0585139	-0.020285	-0.205078
##	1223	-0.293864	0.1227631	0.130031	0.132033
##	1224	0.087327	-0.0011601	-0.095609	0.081453
##	1225	-0.293864	0.1227631	0.130031	0.132033
##	1226	0.031114	-0.1316914	-0.080965	-0.341162
##	1227	-0.293864	0.1227631	0.130031	0.132033
##	1228	-0.293864	0.1227631	0.130031	0.132033
##	1229	-0.293864	0.1227631	0.130031	0.132033
##	1230	-0.293864	0.1227631	0.130031	0.132033
##	1231	-0.110390	-0.0208955	0.010908	-0.135121
##	1232	-0.062345	-0.0585139	-0.020285	-0.205078
##	1233	0.278570	-0.1509019	-0.219776	-0.197013
##	1234	-0.062345	-0.0585139	-0.020285	-0.205078
##	1235	-0.025079	-0.0876930	-0.044481	-0.259341
##	1236	-0.110390	-0.0208955	0.010908	-0.135121
##	1237	0.087327	-0.0011601	-0.095609	0.081453
##	1238	-0.293864	0.1227631	0.130031	0.132033
##	1239	-0.293864	0.1227631	0.130031	0.132033
##	1240	0.005370	-0.1115340	-0.064250	-0.303676
##	1241	-0.293864	0.1227631	0.130031	0.132033
##	1242	-0.293864	0.1227631	0.130031	0.132033
##	1243	-0.293864	0.1227631	0.130031	0.132033
##	1244	-0.293864	0.1227631	0.130031	0.132033
##	1245	0.005370	-0.1115340	-0.064250	-0.303676
##	1246	-0.178105	0.0321246	0.054873	-0.036522
##	1247	-0.293864	0.1227631	0.130031	0.132033
##	1248	-0.110390	-0.0208955	0.010908	-0.135121
##	1249	-0.293864	0.1227631	0.130031	0.132033
##	1250	-0.293864	0.1227631	0.130031	0.132033
##	1251	-0.293864	0.1227631	0.130031	0.132033
##	1252	-0.025079	-0.0876930	-0.044481	-0.259341
##	1253	0.005370	-0.1115340	-0.064250	-0.303676
##	1254	-0.025079	-0.0876930	-0.044481	-0.259341
##	1255	-0.110390	-0.0208955	0.010908	-0.135121
##	1256	0.087327	-0.0011601	-0.095609	0.081453
##	1257	0.005370	-0.1115340	-0.064250	-0.303676
##	1258	-0.293864	0.1227631	0.130031	0.132033
##	1259	-0.293864	0.1227631	0.130031	0.132033
##	1260	0.005370	-0.1115340	-0.064250	-0.303676
##	1261	0.177342	-0.0716413	-0.154053	-0.049617
##	1262	-0.293864	0.1227631	0.130031	0.132033
##	1263	-0.025079	-0.0876930	-0.044481	-0.259341
##	1264	-0.062345	-0.0585139	-0.020285	-0.205078
##	1265	-0.293864	0.1227631	0.130031	0.132033

```
## 1266 -0.110390 -0.0208955 0.010908 -0.135121
## 1267 -0.293864 0.1227631 0.130031 0.132033
## 1268 0.005370 -0.1115340 -0.064250 -0.303676
## 1269 -0.025079 -0.0876930 -0.044481 -0.259341
## 1270 -0.293864 0.1227631 0.130031 0.132033
## 1271 0.129298 -0.0340229 -0.122859 0.020340
## 1272 0.005370 -0.1115340 -0.064250 -0.303676
## 1273 -0.110390 -0.0208955 0.010908 -0.135121
## 1274 0.382197 0.0446073 -0.252422 0.451657
## 1275 0.310299 -0.0012002 -0.218524 0.261252
## 1276 0.277939 -0.0483055 -0.206584 0.098999
## 1277 -0.293864 0.1227631 0.130031 0.132033
## 1278 -0.110390 -0.0208955 0.010908 -0.135121
## 1279 -0.025079 -0.0876930 -0.044481 -0.259341
## 1280 0.301270 0.0058698 -0.212662 0.274400
## 1281 -0.025079 -0.0876930 -0.044481 -0.259341
## 1282 0.005370 -0.1115340 -0.064250 -0.303676
## 1283 -0.293864 0.1227631 0.130031 0.132033
## 1284 0.109627 -0.0186212 -0.110088 0.048982
## 1285 -0.025079 -0.0876930 -0.044481 -0.259341
## 1286 -0.293864 0.1227631 0.130031 0.132033
## 1287 -0.178105 0.0321246 0.054873 -0.036522
## 1288 0.031114 -0.1316914 -0.080965 -0.341162
## 1289 0.146893 -0.0478003 -0.134283 -0.005281
```

```
# have the scores reversed to have the adult lifespan to the left
test.pca_dipper1_reverse <- test.pca_dipper1
test.pca_dipper1_reverse$CA$u <- -(test.pca_dipper1_reverse$CA$u)
test.pca_dipper1_reverse$CA$v <- -(test.pca_dipper1_reverse$CA$v)
g <- vegan::scores(test.pca_dipper1_reverse, display = "species", scaling=2)

# Figure
par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
plot(test.pca_dipper1_reverse, scaling = scl, choices=c(1,2), type="n", ylim=c(-3,3), xlim=c(-3.5,3.5), main="", xlab="PC1 = 34%", ylab="PC2 = 25%", cex.lab=2, cex.axis=2)
with(data_dipper, points(test.pca_dipper1_reverse, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_dipper1_reverse, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1]+0.5, y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2]-0.5, labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```
# Test for proportion of variance explained per axe
sum_all_dipper <- summary(test.pca_dipper1)

prop_dipper <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_dipper[j,i+1] <- as.numeric(abs(sum_all_dipper$species[i,j])/sum(abs(sum_all_dipper$species[i,])))
  }
}

prop_dipper
```

```
##      PC      prim      adult      breed      freq
## 1  1 0.19523693 0.43309730 0.2334169 0.3582304
## 2  2 0.26982551 0.07329361 0.4135437 0.2046272
## 3  3 0.46157243 0.16293131 0.2012361 0.1675604
## 4  4 0.07336513 0.33067778 0.1518033 0.2695820
```

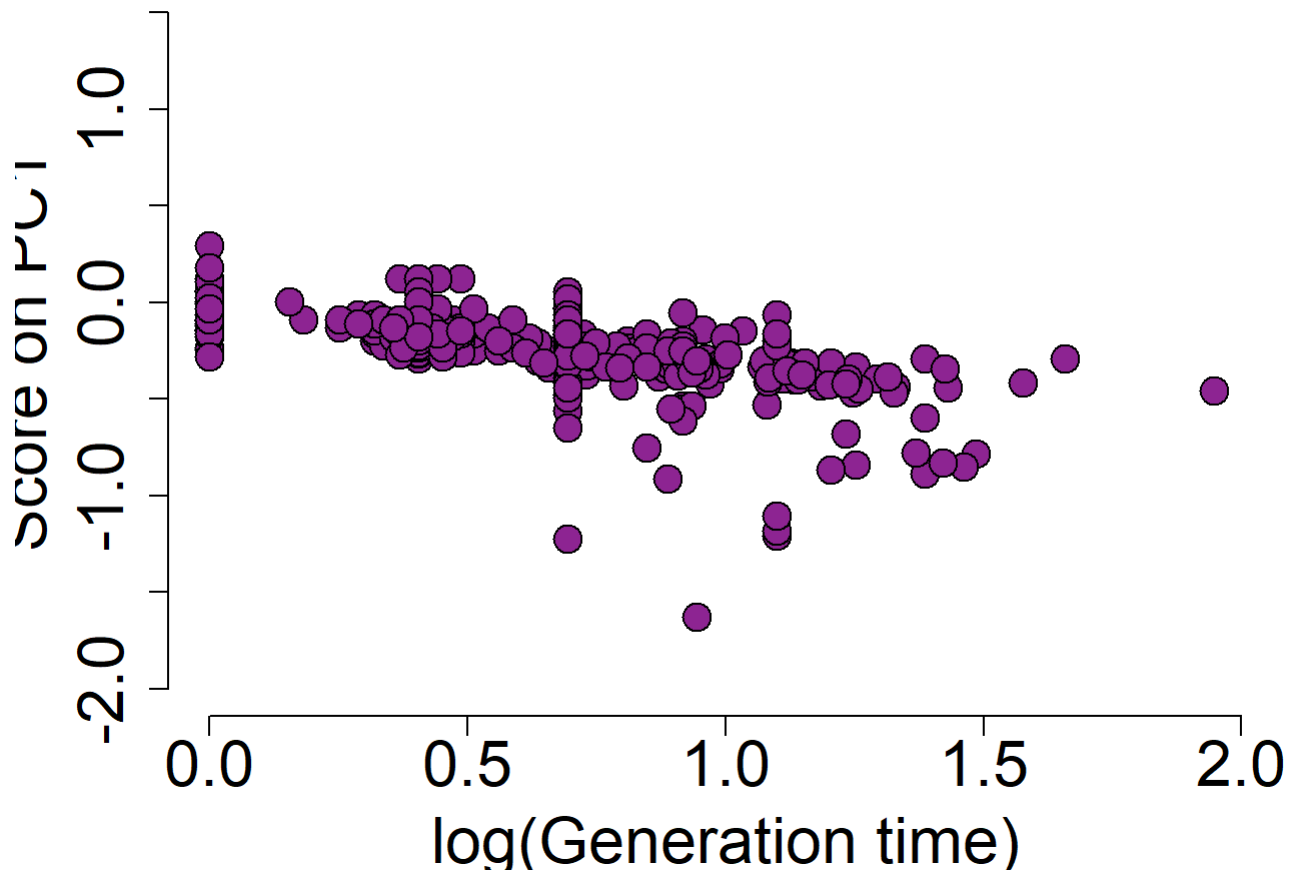
```
# Covariances
test <- as.matrix(scale(log(data_dipper[,c(2,3,4,7)])))
cov(test)
```

```
##           age_first_repro lifetime_adult breed_freq fec_one
## age_first_repro      1.00000000      0.04204146 -0.05099790 0.09384486
## lifetime_adult      0.04204146      1.00000000 -0.15704439 0.26233791
## breed_freq          -0.05099790     -0.15704439  1.00000000 -0.02215099
## fec_one             0.09384486      0.26233791 -0.02215099  1.00000000
```

```
# Correlation with generation time
individual_score_dipper_reverse <- as.numeric(vegan::scores(test.pca_dipper1_reverse)$sites[,1])
summary(lm(individual_score_dipper_reverse ~ log(data_dipper$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_dipper_reverse ~ log(data_dipper$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.27746 -0.09312 -0.03743  0.17634  0.41539
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.117528   0.004929   23.85  <2e-16 ***
## log(data_dipper$generation_time) -0.497577   0.011269  -44.15  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1489 on 1287 degrees of freedom
## Multiple R-squared:  0.6023, Adjusted R-squared:  0.602
## F-statistic: 1949 on 1 and 1287 DF, p-value: < 2.2e-16
```

```
par(bty="n")
plot(log(data_dipper$generation_time), individual_score_dipper_reverse, pch=21, xlab="log(Generation time)", ylab="Score on PC1", xlim=c(0,2), ylim=c(-2,1.5), bg=colors[5], cex=2, cex.axis=2, cex.lab=2)
```



House sparrow

```
data_sparrow <- data[which(data$species == "housesparrow"),]  
data_sparrow$fec_one <- data_sparrow$fecundity + 1  
test.pca_sparrow1 <- rda(log(data_sparrow[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_sparrow1)
```

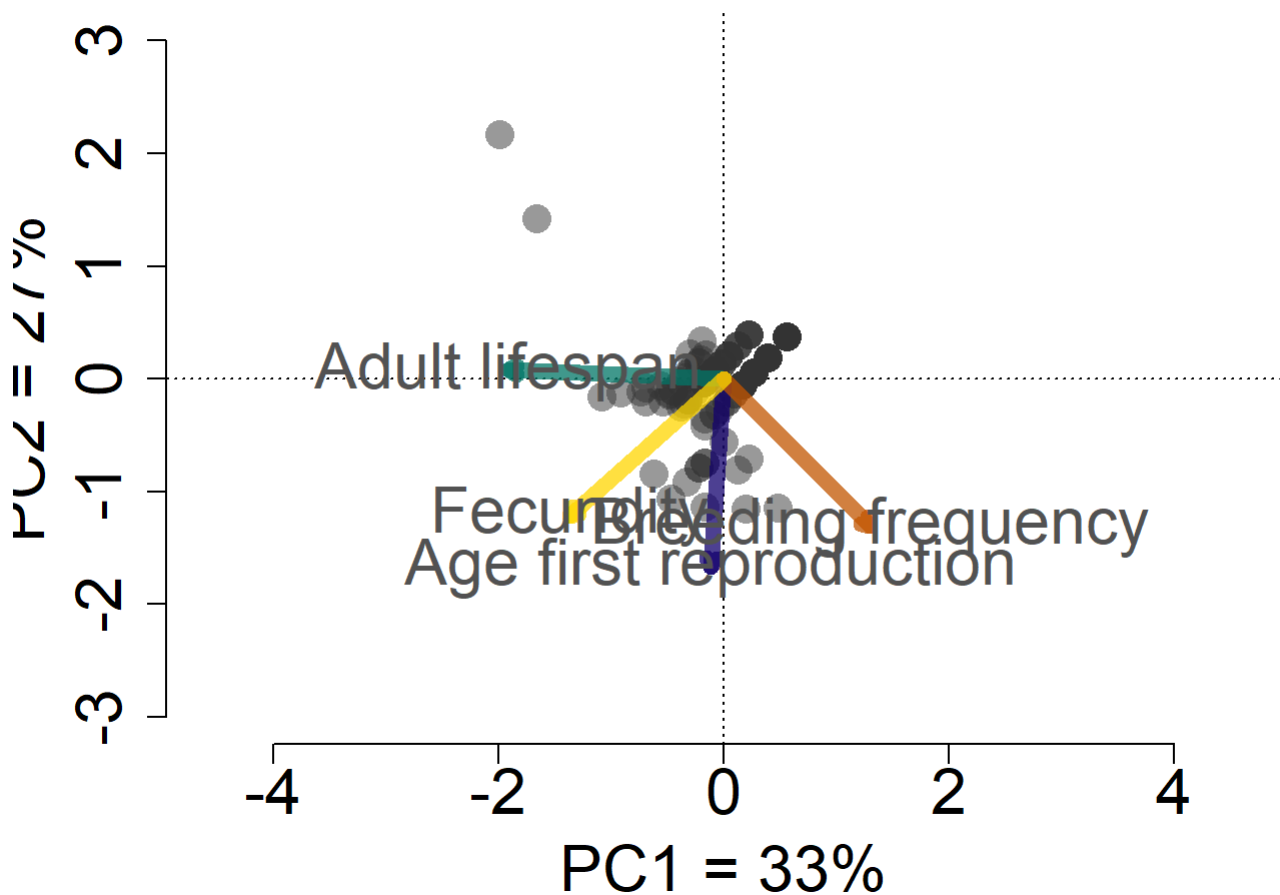
```
##
## Call:
## rda(X = log(data_sparrow[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained    4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      1.3038 1.0770 0.9316 0.6876
## Proportion Explained 0.3259 0.2693 0.2329 0.1719
## Cumulative Proportion 0.3259 0.5952 0.8281 1.0000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 4.709679
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro -0.1114 -1.6690 -1.65711 0.03481
## lifetime_adult -1.9128 0.0792 0.06188 1.36985
## breed_freq      1.3000 -1.3048 1.10191 0.96873
## fec_one         -1.3670 -1.2157 1.09637 -0.99833
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 1327  -0.470637 -1.06659 -0.33843 -0.29038
## 2126  -0.280607 0.08178 0.15255 0.44295
## 3123  -0.320246 -0.18814 0.41430 -0.16517
## 4124  -0.916203 -0.12757 0.41194 0.71885
## 5120  -0.191267 0.33920 -0.10188 0.95919
## 6120  -0.175137 -0.03193 0.25143 0.03577
## 7117  -0.377834 -0.25013 0.47894 -0.24492
## 8117   0.564844 0.37620 -0.20772 0.11475
## 9109   0.480830 -1.14734 -1.95649 0.16452
## 10116  0.564844 0.37620 -0.20772 0.11475
## 11109 -1.660944 1.42241 -1.13751 -1.10882
## 12105 -0.129699 0.01699 0.20043 0.09870
## 1328  -0.156539 0.21535 0.01329 0.61476
## 1426   0.265912 0.05439 0.12780 -0.29921
## 1527  -0.078903 0.07167 0.14342 0.16904
## 1626  -0.217144 -0.79369 -0.62295 0.06066
## 1727   0.564844 0.37620 -0.20772 0.11475
```

##	1827	-0.166793	-0.35100	-1.05088	1.07613
##	1925	0.564844	0.37620	-0.20772	0.11475
##	2025	-0.223019	0.14378	0.08791	0.52269
##	2127	0.564844	0.37620	-0.20772	0.11475
##	2225	0.564844	0.37620	-0.20772	0.11475
##	2325	0.265912	0.05439	0.12780	-0.29921
##	2425	0.564844	0.37620	-0.20772	0.11475
##	2525	-0.021315	0.13367	0.07878	0.24879
##	2623	0.564844	0.37620	-0.20772	0.11475
##	2723	0.169678	-0.04921	0.23582	-0.43248
##	2823	-0.171705	-0.74478	-0.67395	0.12358
##	2923	0.564844	0.37620	-0.20772	0.11475
##	3023	-0.217144	-0.79369	-0.62295	0.06066
##	3124	-0.033019	-0.26742	0.46333	-0.71317
##	3223	0.169678	-0.04921	0.23582	-0.43248
##	3323	0.091049	-0.13385	0.32407	-0.54136
##	3423	0.091049	-0.13385	0.32407	-0.54136
##	3523	0.123794	0.28988	-0.08409	0.44973
##	3620	0.169678	-0.04921	0.23582	-0.43248
##	3720	0.220029	0.39348	-0.19211	0.58300
##	3823	0.003158	-0.55653	-0.87022	0.36573
##	3920	0.265912	0.05439	0.12780	-0.29921
##	4023	0.564844	0.37620	-0.20772	0.11475
##	4125	-0.171705	-0.74478	-0.67395	0.12358
##	4223	0.220029	0.39348	-0.19211	0.58300
##	4323	-1.984404	2.16544	-1.87383	-1.57018
##	4423	-0.315126	0.04462	0.19129	0.39514
##	4523	0.169678	-0.04921	0.23582	-0.43248
##	4620	0.127671	-0.81098	-0.63856	-0.40759
##	4720	0.265912	0.05439	0.12780	-0.29921
##	4820	-0.078903	0.07167	0.14342	0.16904
##	4920	-0.330293	-0.91550	-0.49595	-0.09603
##	5020	0.389980	0.18795	-0.01145	-0.12740
##	5123	-0.561056	-0.05889	0.31318	0.44710
##	5220	-1.078617	-0.16156	0.45957	0.83684
##	5320	0.389980	0.18795	-0.01145	-0.12740
##	5420	-0.320246	-0.18814	0.41430	-0.16517
##	5520	-0.694966	-0.07799	0.34391	0.56613
##	5620	-0.455470	-0.10646	0.34882	0.20080
##	5720	-0.175137	-0.03193	0.25143	0.03577
##	5820	0.389980	0.18795	-0.01145	-0.12740
##	5920	-0.622234	-0.84130	-0.53969	0.44544
##	6020	0.564844	0.37620	-0.20772	0.11475
##	6123	0.389980	0.18795	-0.01145	-0.12740
##	6219	-0.320246	-0.18814	0.41430	-0.16517
##	6319	0.265912	0.05439	0.12780	-0.29921
##	6419	0.265912	0.05439	0.12780	-0.29921
##	6518	0.265912	0.05439	0.12780	-0.29921
##	6618	0.564844	0.37620	-0.20772	0.11475
##	6718	-0.171260	-1.13278	-0.30304	-0.82155
##	6818	0.389980	0.18795	-0.01145	-0.12740
##	6918	-0.223019	0.14378	0.08791	0.52269

## 7017	-0.376841	-0.02182	0.26056	0.30968
## 7118	0.169678	-0.04921	0.23582	-0.43248
## 7217	0.045165	0.20524	0.00416	0.34085
## 7317	0.123794	0.28988	-0.08409	0.44973
## 7417	-0.201977	0.16643	0.06429	0.55183
## 7517	0.265912	0.05439	0.12780	-0.29921
## 7617	-0.299650	0.22252	0.01977	0.80910
## 7717	0.123794	0.28988	-0.08409	0.44973
## 7817	-0.478788	-0.13156	0.37499	0.16851
## 7917	0.024569	-0.20542	0.39869	-0.63342
## 8017	0.265912	0.05439	0.12780	-0.29921
## 8118	-0.694816	-0.20289	0.46331	0.26187
## 8216	-0.489990	-0.14362	0.38756	0.15299
## 8317	-0.732904	-0.11883	0.38649	0.51359
## 8417	-0.542013	-0.19963	0.44595	0.08095
## 8516	0.024569	-0.20542	0.39869	-0.63342
## 8617	0.265912	0.05439	0.12780	-0.29921
## 8716	0.564844	0.37620	-0.20772	0.11475
## 8817	0.169678	-0.04921	0.23582	-0.43248
## 8917	0.265912	0.05439	0.12780	-0.29921
## 9017	0.223905	-0.70738	-0.74658	-0.27432
## 9117	-0.083815	-0.32210	0.52034	-0.78351
## 9216	-0.175137	-0.03193	0.25143	0.03577
## 9316	-0.021315	0.13367	0.07878	0.24879
## 9415	0.564844	0.37620	-0.20772	0.11475
## 9515	-0.347087	0.01022	0.22717	0.35089
## 9615	-0.129699	0.01699	0.20043	0.09870
## 9716	0.265912	0.05439	0.12780	-0.29921
## 9816	-0.170357	-0.41527	0.61748	-0.90336
## 9916	-0.078903	0.07167	0.14342	0.16904
## 10016	0.265912	0.05439	0.12780	-0.29921
## 10117	0.265912	0.05439	0.12780	-0.29921
## 10216	0.169678	-0.04921	0.23582	-0.43248
## 10316	-0.175137	-0.03193	0.25143	0.03577
## 10416	0.564844	0.37620	-0.20772	0.11475
## 10516	-0.021315	0.13367	0.07878	0.24879
## 10616	0.045165	0.20524	0.00416	0.34085
## 10716	0.220029	0.39348	-0.19211	0.58300
## 10816	0.199333	-1.15298	-1.25806	-0.25977
## 10915	-0.078903	0.07167	0.14342	0.16904
## 11015	0.045165	0.20524	0.00416	0.34085
## 11115	-0.253766	-0.11657	0.33969	-0.07311
## 11215	-0.455470	-0.10646	0.34882	0.20080
## 11314	-0.078903	0.07167	0.14342	0.16904
## 11415	0.220029	0.39348	-0.19211	0.58300
## 11515	0.169678	-0.04921	0.23582	-0.43248
## 11615	-0.083815	-0.32210	0.52034	-0.78351
## 11715	0.265912	0.05439	0.12780	-0.29921
## 11815	-0.175137	-0.03193	0.25143	0.03577
## 11915	0.389980	0.18795	-0.01145	-0.12740
## 12015	0.389980	0.18795	-0.01145	-0.12740
## 12115	0.389980	0.18795	-0.01145	-0.12740

```
## 12215  0.045165  0.20524  0.00416  0.34085
## 12315 -0.350001 -0.22017  0.44770 -0.20638
## 12414  0.169678 -0.04921  0.23582 -0.43248
```

```
# Figure
par(mar=c(4,4,2,2), bty="n")
plot(test.pca_sparrow1, scaling = scl, choices=c(1,2), type="n", main="", ylim=c(-3,3), xlim=c(-3.5,3.5), xlab="PC1 = 33%", ylab="PC2 = 27%", cex.lab=2, cex.axis=2)
with(data_sparrow, points(test.pca_sparrow1, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_sparrow1, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```

# Test for proportion of variance explained per axe
sum_all_sparrow <- summary(test.pca_sparrow1)

prop_sparrow <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_sparrow[j,i+1] <- as.numeric(abs(sum_all_sparrow$species[i,j])/sum(abs(sum_all_sparrow
$species[i,])))
  }
}

prop_sparrow

```

```

##      PC      prim      adult      breed      freq
## 1  1 0.03208581 0.55868552 0.2780457 0.2922582
## 2  2 0.48066141 0.02313258 0.2790827 0.2599057
## 3  3 0.47722855 0.01807415 0.2356775 0.2343976
## 4  4 0.01002423 0.40010775 0.2071941 0.2134385

```

```

# Covariances
test <- as.matrix(scale(log(data_sparrow[,c(2,3,4,7)])))
cov(test)

```

```

##           age_first_repro lifetime_adult  breed_freq  fec_one
## age_first_repro      1.000000000      0.004699795  0.043411674 0.059466064
## lifetime_adult      0.004699795      1.000000000 -0.215451747 0.219784074
## breed_freq          0.043411674     -0.215451747  1.000000000 0.009043238
## fec_one             0.059466064      0.219784074  0.009043238 1.000000000

```

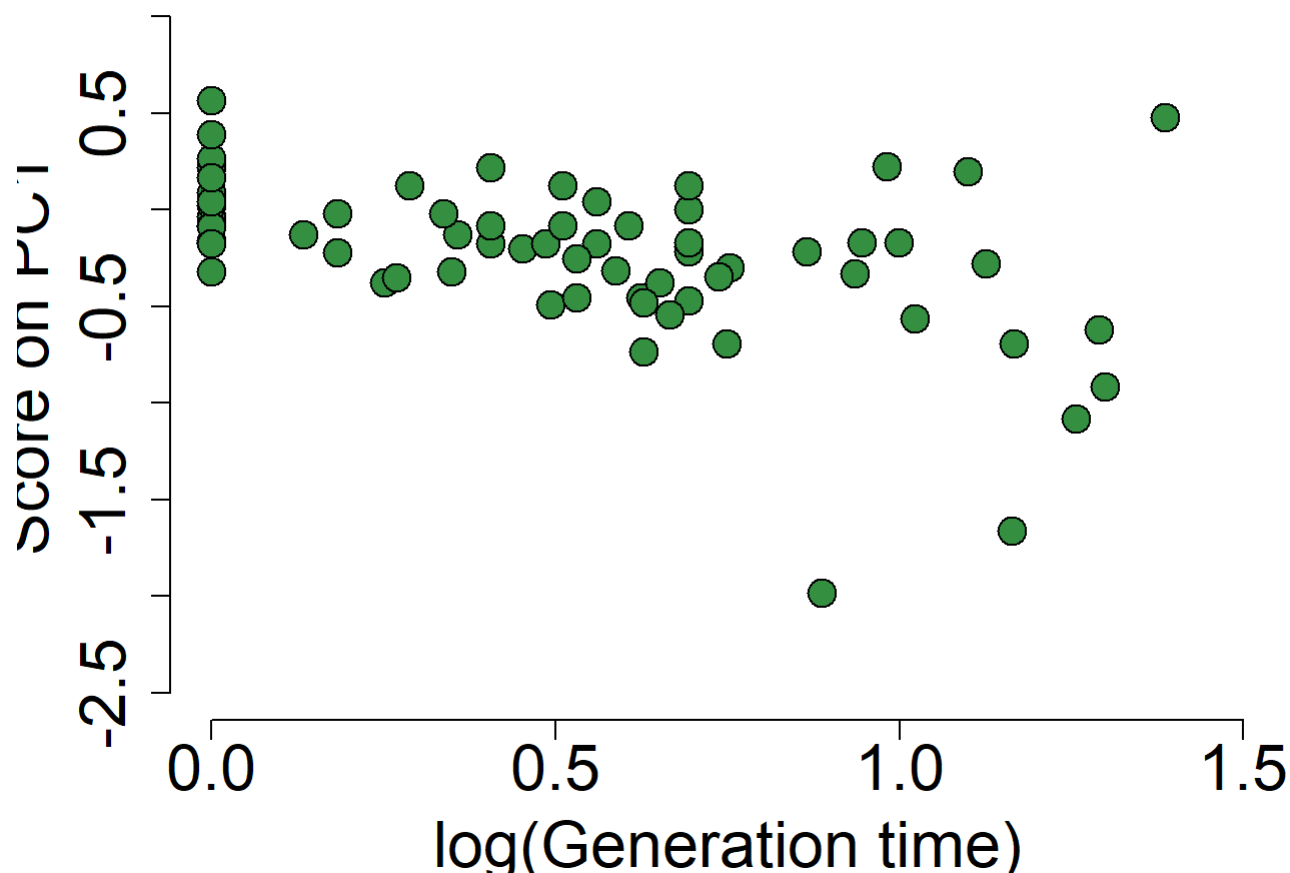
```

# Correlation with generation time
individual_score_sparrow <- as.numeric(vegan::scores(test.pca_sparrow1)$sites[,1])
summary(lm(individual_score_sparrow ~ log(data_sparrow$generation_time)))

```

```
##
## Call:
## lm(formula = individual_score_sparrow ~ log(data_sparrow$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.57081 -0.17260  0.01211  0.17222  1.24947
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.21776    0.03674   5.927 2.93e-08 ***
## log(data_sparrow$generation_time) -0.71154    0.07412  -9.601 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3218 on 122 degrees of freedom
## Multiple R-squared:  0.4304, Adjusted R-squared:  0.4257
## F-statistic: 92.17 on 1 and 122 DF,  p-value: < 2.2e-16
```

```
par(bty="n")
plot(log(data_sparrow$generation_time), individual_score_sparrow, pch=21, xlab="log(Generation t
ime)", ylab="Score on PC1", xlim=c(0,1.5), ylim=c(-2.5,1), bg=colors[8], cex=2, cex.axis=2, cex.
lab=2)
```



Blue tits

```
data_btits <- data[which(data$species == "bluetits"),]  
data_btits$fec_one <- data_btits$fecundity + 1  
test.pca_btits1 <- rda(log(data_btits[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_btits1)
```

```
##
## Call:
## rda(X = log(data_btits[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           3           1
## Unconstrained   3           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3
## Eigenvalue      1.1615 1.0246 0.8139
## Proportion Explained 0.3872 0.3415 0.2713
## Cumulative Proportion 0.3872 0.7287 1.0000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 5.174636
##
##
## Species scores
##
##           PC1    PC2    PC3
## age_first_repro 0.05488 2.8415 -0.9211
## lifetime_adult 2.28455 0.6939 1.7958
## breed_freq      0.00000 0.0000 0.0000
## fec_one         2.26830 -0.7676 -1.7864
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3
## 1329 -0.104166 -0.096594 -0.234119
## 2128 0.065252 0.077571 0.313277
## 3125 -0.263287 -0.035550 -0.055279
## 4126 -0.376185 0.007760 0.071609
## 5124 0.440014 0.172960 0.629882
## 6124 0.111475 0.059839 0.261326
## 7119 0.653350 0.091118 0.390108
## 8119 -0.047646 0.120882 0.440165
## 9118 0.111475 0.059839 0.261326
## 10118 0.288847 0.131692 0.493564
## 11116 -0.047646 0.120882 0.440165
## 12106 -0.104166 -0.096594 -0.234119
## 1330 0.008732 -0.139905 -0.361007
## 1427 0.477051 0.059492 0.282037
## 1528 0.065252 0.077571 0.313277
## 1627 0.008732 -0.139905 -0.361007
## 1728 -0.104166 -0.096594 -0.234119
```

## 1828	-0.535306	0.068804	0.250449
## 1926	0.437774	0.074560	0.326181
## 2026	-0.263287	-0.035550	-0.055279
## 2129	0.536210	0.136056	0.521765
## 2226	0.370333	0.100432	0.401980
## 2326	0.730228	0.138617	0.541225
## 2426	0.284868	-0.006680	0.066446
## 2526	0.054955	-0.157637	-0.412958
## 2624	0.627178	0.101158	0.419524
## 2724	-0.043671	-0.119801	-0.302110
## 2824	-0.043671	-0.119801	-0.302110
## 2924	0.311943	-0.017067	0.036015
## 3024	0.190226	0.029627	0.172816
## 3126	0.190226	0.029627	0.172816
## 3224	-0.104166	-0.096594	-0.234119
## 3324	0.255785	0.004477	0.099133
## 3424	-0.043671	-0.119801	-0.302110
## 3524	-0.807325	0.173158	0.556176
## 3623	0.008732	-0.139905	-0.361007
## 3723	0.337271	-0.026783	0.007549
## 3824	-0.043671	-0.119801	-0.302110
## 3923	0.223288	0.156842	0.567246
## 4024	0.008732	-0.139905	-0.361007
## 4127	0.284868	-0.006680	0.066446
## 4224	0.008732	-0.139905	-0.361007
## 4324	-0.376185	0.007760	0.071609
## 4424	-0.807325	0.173158	0.556176
## 4524	0.284868	-0.006680	0.066446
## 4623	-0.175716	-0.069145	-0.153702
## 4723	-0.807325	0.173158	0.556176
## 4823	0.394124	0.091305	0.375240
## 4923	0.054955	-0.157637	-0.412958
## 5023	0.008732	-0.139905	-0.361007
## 5125	-0.175716	-0.069145	-0.153702
## 5223	0.255785	0.004477	0.099133
## 5323	0.152823	0.043976	0.214854
## 5423	-0.807325	0.173158	0.556176
## 5523	0.223288	0.156842	0.567246
## 5623	-0.807325	0.173158	0.556176
## 5723	-0.807325	0.173158	0.556176
## 5823	0.008732	-0.139905	-0.361007
## 5923	-0.104166	-0.096594	-0.234119
## 6023	0.224373	0.016528	0.134437
## 6125	-0.807325	0.173158	0.556176
## 6220	-0.807325	0.173158	0.556176
## 6320	0.008732	-0.139905	-0.361007
## 6420	0.394124	0.091305	0.375240
## 6519	-0.807325	0.173158	0.556176
## 6619	-0.043671	-0.119801	-0.302110
## 6719	-0.175716	-0.069145	-0.153702
## 6819	0.383494	-0.044516	-0.044402
## 6919	-0.175716	-0.069145	-0.153702

## 7018	-0.043671	-0.119801	-0.302110
## 7120	0.568932	0.123503	0.484988
## 7218	-0.104166	-0.096594	-0.234119
## 7318	0.096303	-0.173499	-0.459429
## 7418	0.190226	0.029627	0.172816
## 7518	-0.104166	-0.096594	-0.234119
## 7618	-0.175716	-0.069145	-0.153702
## 7718	-0.263287	-0.035550	-0.055279
## 7818	-0.104166	-0.096594	-0.234119
## 7918	-0.376185	0.007760	0.071609
## 8018	-0.535306	0.068804	0.250449
## 8120	0.552912	0.129649	0.502993
## 8217	-0.175716	-0.069145	-0.153702
## 8318	-0.535306	0.068804	0.250449
## 8418	-0.175716	-0.069145	-0.153702
## 8517	0.152823	0.043976	0.214854
## 8618	-0.043671	-0.119801	-0.302110
## 8717	-0.175716	-0.069145	-0.153702
## 8818	0.094852	4.911393	-1.592013
## 8918	-0.043671	-0.119801	-0.302110
## 9018	-0.175716	-0.069145	-0.153702
## 9119	-0.263287	-0.035550	-0.055279
## 9217	-0.175716	-0.069145	-0.153702
## 9317	-0.206767	0.181925	0.619005
## 9416	0.370333	0.100432	0.401980
## 9516	-0.043671	-0.119801	-0.302110
## 9616	0.012849	0.097674	0.372174
## 9717	0.345005	0.110148	0.430446
## 9817	-0.263287	-0.035550	-0.055279
## 9917	-0.535306	0.068804	0.250449
## 10017	0.255785	0.004477	0.099133
## 10119	0.337271	-0.026783	0.007549
## 10217	0.054955	-0.157637	-0.412958
## 10317	0.111475	0.059839	0.261326
## 10417	0.008732	-0.139905	-0.361007
## 10517	-0.535306	0.068804	0.250449
## 10617	-0.175716	-0.069145	-0.153702
## 10717	0.224373	0.016528	0.134437
## 10817	-0.175716	-0.069145	-0.153702
## 10916	-0.043671	-0.119801	-0.302110
## 11016	0.311943	-0.017067	0.036015
## 11117	0.054955	-0.157637	-0.412958
## 11216	0.054955	-0.157637	-0.412958
## 11315	-0.104166	-0.096594	-0.234119
## 11416	0.255785	0.004477	0.099133
## 11516	-0.104166	-0.096594	-0.234119
## 11616	-0.175716	-0.069145	-0.153702
## 11716	0.152823	0.043976	0.214854
## 11816	-0.104166	-0.096594	-0.234119
## 11916	-0.104166	-0.096594	-0.234119
## 12016	0.284868	-0.006680	0.066446
## 12116	0.255785	0.004477	0.099133

12216 -0.175716 -0.069145 -0.153702
12316 0.054955 -0.157637 -0.412958
12415 -0.263287 -0.035550 -0.055279
12515 0.560866 0.027338 0.187836
12615 0.361063 -0.035910 -0.019191
12715 -0.104166 -0.096594 -0.234119
12815 0.008732 -0.139905 -0.361007
12913 0.255785 0.004477 0.099133
13010 0.284868 -0.006680 0.066446
13112 -0.104166 -0.096594 -0.234119
13210 0.311943 -0.017067 0.036015
1336 0.054955 -0.157637 -0.412958
1346 -0.175716 -0.069145 -0.153702
1356 -0.043671 -0.119801 -0.302110
1366 -0.043671 -0.119801 -0.302110
1376 -0.104166 -0.096594 -0.234119
1386 0.008732 -0.139905 -0.361007
1396 0.008732 -0.139905 -0.361007
1406 -0.043671 -0.119801 -0.302110
14112 0.008732 -0.139905 -0.361007
1428 0.054955 -0.157637 -0.412958
1436 0.008732 -0.139905 -0.361007
1446 -0.104166 -0.096594 -0.234119
1456 -0.807325 0.173158 0.556176
1466 0.416556 0.082699 0.350029
1476 -0.807325 0.173158 0.556176
1486 0.512752 0.045796 0.241912
1496 0.512752 0.045796 0.241912
1506 -0.263287 -0.035550 -0.055279
15112 -0.807325 0.173158 0.556176
1529 -0.104166 -0.096594 -0.234119
1536 -0.807325 0.173158 0.556176
1546 -0.104166 -0.096594 -0.234119
1556 0.337271 -0.026783 0.007549
1566 0.734039 0.200062 0.731010
1576 0.152823 0.043976 0.214854
1586 0.255785 0.004477 0.099133
1595 -0.104166 -0.096594 -0.234119
1606 -0.807325 0.173158 0.556176
16112 -0.175716 -0.069145 -0.153702
1628 0.008732 -0.139905 -0.361007
1635 -0.104166 -0.096594 -0.234119
1646 -0.104166 -0.096594 -0.234119
1656 0.512752 0.045796 0.241912
1666 0.500509 0.149752 0.561890
1676 -0.043671 -0.119801 -0.302110
1686 0.311943 -0.017067 0.036015
1696 -0.263287 -0.035550 -0.055279
1706 0.284868 -0.006680 0.066446
17112 -0.807325 0.173158 0.556176
1729 0.152823 0.043976 0.214854
1735 -0.807325 0.173158 0.556176

## 1746	-0.043671	-0.119801	-0.302110
## 1756	0.416556	0.082699	0.350029
## 1766	0.008732	-0.139905	-0.361007
## 1776	0.613407	0.106442	0.435002
## 1786	0.008732	-0.139905	-0.361007
## 1796	0.008732	-0.139905	-0.361007
## 1805	-0.104166	-0.096594	-0.234119
## 18112	-0.104166	-0.096594	-0.234119
## 1829	0.394124	0.091305	0.375240
## 1836	-0.104166	-0.096594	-0.234119
## 1846	-0.263287	-0.035550	-0.055279
## 1856	-0.807325	0.173158	0.556176
## 1866	0.599134	0.111917	0.451043
## 1876	-0.807325	0.173158	0.556176
## 1885	0.008732	-0.139905	-0.361007
## 1895	0.529454	0.039389	0.223140
## 1905	0.111475	0.059839	0.261326
## 19112	-0.043671	-0.119801	-0.302110
## 1927	-0.104166	-0.096594	-0.234119
## 1935	0.477051	0.059492	0.282037
## 1945	0.311943	-0.017067	0.036015
## 1955	-0.043671	-0.119801	-0.302110
## 1965	-0.104166	-0.096594	-0.234119
## 1975	0.311943	-0.017067	0.036015
## 1985	-0.104166	-0.096594	-0.234119
## 1995	0.054955	-0.157637	-0.412958
## 2005	0.416556	0.082699	0.350029
## 20112	0.640482	0.096055	0.404571
## 2027	0.152823	0.043976	0.214854
## 2035	-0.263287	-0.035550	-0.055279
## 2045	-0.175716	-0.069145	-0.153702
## 2055	0.190226	0.029627	0.172816
## 2065	0.255785	0.004477	0.099133
## 2075	0.096303	-0.173499	-0.459429
## 2085	0.223288	0.156842	0.567246
## 2095	0.311943	-0.017067	0.036015
## 21011	0.152823	0.043976	0.214854
## 21112	0.008732	-0.139905	-0.361007
## 21210	-0.043671	-0.119801	-0.302110
## 2135	-0.175716	-0.069145	-0.153702
## 2145	0.284868	-0.006680	0.066446
## 2155	0.284868	-0.006680	0.066446
## 2165	0.255785	0.004477	0.099133
## 2175	-0.175716	-0.069145	-0.153702
## 2185	-0.043671	-0.119801	-0.302110
## 2195	0.284868	-0.006680	0.066446
## 2205	0.008732	-0.139905	-0.361007
## 22112	0.345005	0.110148	0.430446
## 2227	-0.043671	-0.119801	-0.302110
## 2235	0.190226	0.029627	0.172816
## 2245	0.190226	0.029627	0.172816
## 2255	-0.263287	-0.035550	-0.055279

```
## 2265 -0.043671 -0.119801 -0.302110
## 2275 0.111475 0.059839 0.261326
## 2285 -0.104166 -0.096594 -0.234119
## 2295 -0.104166 -0.096594 -0.234119
## 2305 0.008732 -0.139905 -0.361007
## 23112 0.008732 -0.139905 -0.361007
## 2327 0.008732 -0.139905 -0.361007
## 2335 0.008732 -0.139905 -0.361007
## 2345 0.008732 -0.139905 -0.361007
## 2355 -0.043671 -0.119801 -0.302110
## 2365 0.008732 -0.139905 -0.361007
## 2375 -0.043671 -0.119801 -0.302110
## 2385 -0.104166 -0.096594 -0.234119
## 2395 -0.043671 -0.119801 -0.302110
## 2405 -0.175716 -0.069145 -0.153702
```

```
# have the scores reversed to have the adult lifespan to the left
```

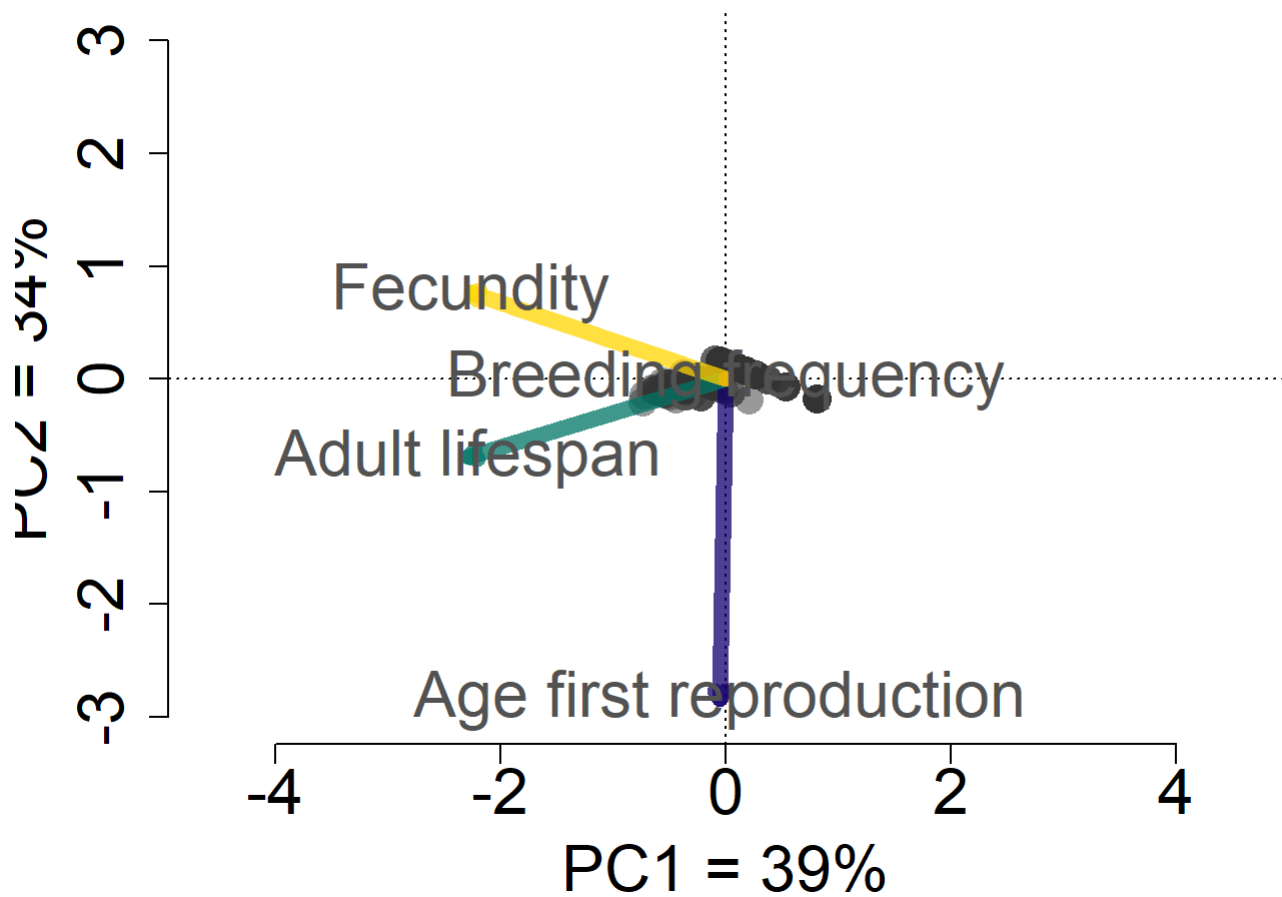
```
test.pca_btits1_reverse <- test.pca_btits1
test.pca_btits1_reverse$CA$u <- -(test.pca_btits1_reverse$CA$u)
test.pca_btits1_reverse$CA$v <- -(test.pca_btits1_reverse$CA$v)
g <- vegan::scores(test.pca_btits1_reverse, display = "species", scaling=2)
```

```
# Figure
```

```
par(mar=c(4,4,2,2), bty="n")
plot(test.pca_btits1_reverse, scaling = scl, choices=c(1,2), type="n", main="", ylim=c(-3,3), x
lim=c(-3.5,3.5), xlab="PC1 = 39%", ylab="PC2 = 34%", cex.lab=2, cex.axis=2)
with(data_btits, points(test.pca_btits1_reverse, display = "sites", col = col_points, scaling =
scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_btits1_reverse, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, c
ol_fec), lwd=8)
```

```
## Warning in arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col =
## c(col_afr, : zero-length arrow is of indeterminate angle and so skipped
```

```
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```
# Test for proportion of variance explained per axe
sum_all_btits <- summary(test.pca_btits1)

prop_btits <- data.frame(
  PC = c(1:3),
  prim = c(1:3),
  adult = c(1:3),
  breed = c(1:3),
  freq = c(1:3)
)

for (i in 1:4) {
  for (j in 1:3){
    prop_btits[j,i+1] <- as.numeric(abs(sum_all_btits$species[i,j])/sum(abs(sum_all_btits$species[i,])))
  }
}

prop_btits
```

```
##      PC      prim      adult breed      freq
## 1  1 0.01437523 0.4785170   NaN 0.4703781
## 2  2 0.74434692 0.1453326   NaN 0.1591714
## 3  3 0.24127785 0.3761504   NaN 0.3704505
```

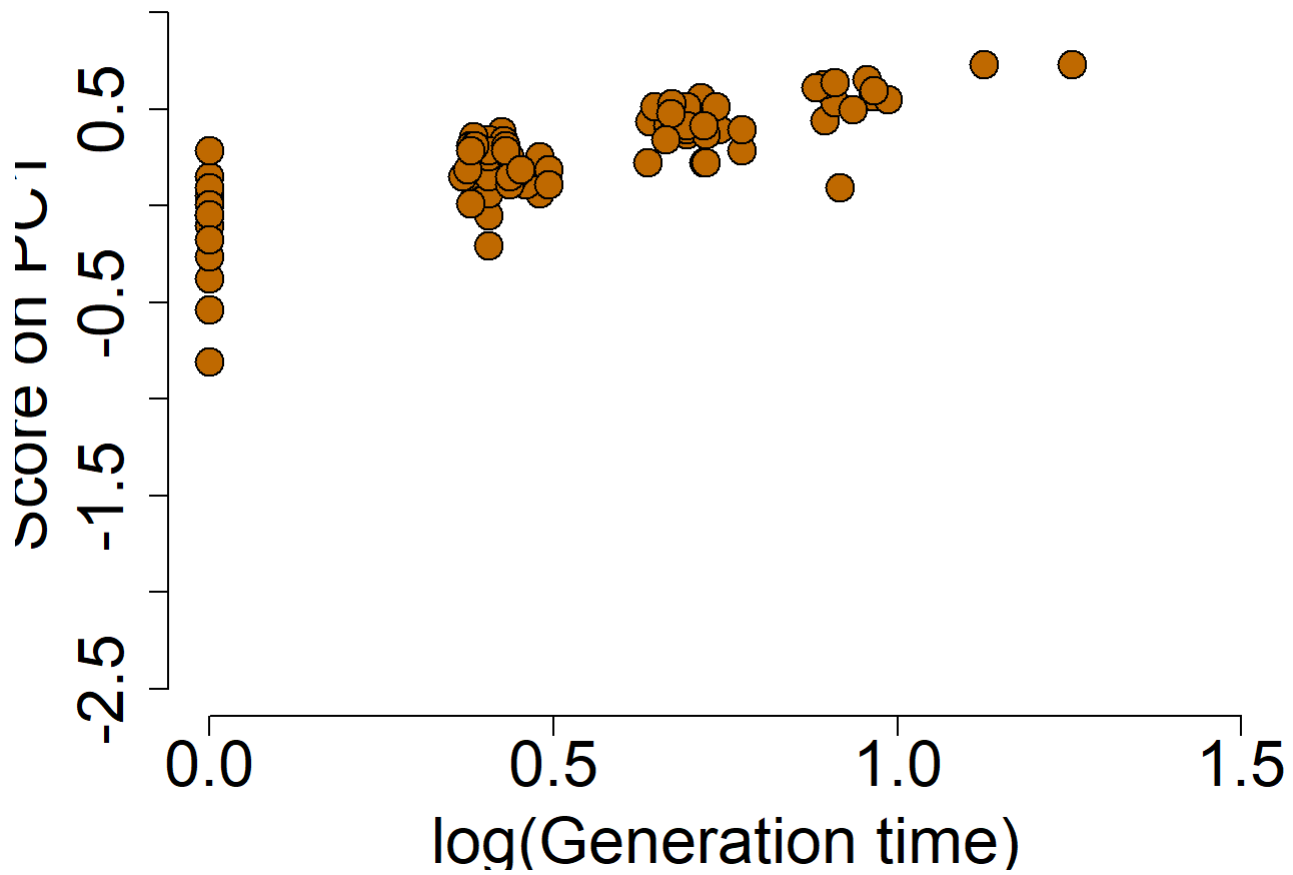
```
# Covariances
test <- as.matrix(scale(log(data_btits[,c(2,3,4,7)])))
cov(test)
```

```
##              age_first_repro lifetime_adult breed_freq   fec_one
## age_first_repro      1.00000000      0.04961868      NA -0.0460664
## lifetime_adult      0.04961868      1.00000000      NA  0.1614858
## breed_freq              NA              NA      NA      NA
## fec_one              -0.04606640      0.16148583      NA  1.0000000
```

```
# Correlation with generation time
individual_score_btits <- as.numeric(vegan::scores(test.pca_btits1)$sites[,1])
summary(lm(individual_score_btits ~ log(data_btits$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_btits ~ log(data_btits$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.64009 -0.04288  0.06307  0.12418  0.45210
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -0.16724     0.01764   -9.481  <2e-16 ***
## log(data_btits$generation_time)  0.81981     0.04853  16.894  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2262 on 238 degrees of freedom
## Multiple R-squared:  0.5453, Adjusted R-squared:  0.5434
## F-statistic: 285.4 on 1 and 238 DF,  p-value: < 2.2e-16
```

```
par(bty="n")
plot(log(data_btits$generation_time), individual_score_btits, pch=21, xlab="log(Generation time)",
     ylab="Score on PC1", xlim=c(0,1.5), ylim=c(-2.5,1), bg=colors[4], cex=2, cex.axis=2, cex.lab=2)
```



Red-Billed gull

```
data_gull <- data[which(data$species == "redbilledgull"),]  
data_gull$fec_one <- data_gull$fecundity + 1  
test.pca_gull1 <- rda(log(data_gull[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_gull1)
```

```
##
## Call:
## rda(X = log(data_gull[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained   4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      1.5620 1.2103 0.7970 0.4308
## Proportion Explained 0.3905 0.3026 0.1992 0.1077
## Cumulative Proportion 0.3905 0.6931 0.8923 1.0000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 7.21643
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro 1.4191 -2.3456 2.2897 0.5110
## lifetime_adult -0.2279 3.0252 1.7939 0.7727
## breed_freq      -3.0424 -0.9979 -0.3798 1.6196
## fec_one         -3.0023 -0.3272 1.3309 -1.4584
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 1360  0.5153689 0.5777575 0.1715935 -5.016e-01
## 2149 -0.4200862 -0.0185366 0.0438268 -6.826e-02
## 3136  0.1455123 0.3390855 0.0242793 9.015e-02
## 4137 -0.2011384 -0.0179012 0.0744151 1.227e-01
## 5135 -0.2530902 0.0224110 0.2962613 2.205e-01
## 6135  0.2261179 -0.0267863 0.4100324 2.903e-01
## 7128  0.0093645 0.2877259 0.0682028 3.568e-02
## 8128 -0.1562925 0.0839678 -0.3825609 3.772e-01
## 9127  0.0005398 -0.1278434 0.2647115 1.709e-01
## 10127 0.0665967 -0.8699872 -0.6818693 2.638e-01
## 11124 -0.0691898 -0.2613080 -0.1301734 3.182e-01
## 12120 0.1035027 0.0262560 0.3043146 -1.436e-01
## 1370  0.0531113 0.3816502 -0.1812526 2.600e-01
## 1440 -0.0555391 0.1488989 -0.2701482 -1.083e-01
## 1550 -0.0565968 -0.0865003 0.0442581 3.041e-01
## 1640 -0.0998082 -0.0220235 0.1826037 1.001e-01
## 1740  0.2410832 0.0978768 0.0921243 1.092e-01
```

##	1850	-0.1211563	-0.1365624	0.2374754	3.243e-02
##	1938	0.1625321	-0.0799831	-0.0250799	3.609e-02
##	2038	-0.0066788	-0.2414075	0.3524219	2.098e-01
##	2150	-0.4814030	0.0824977	0.3902453	-5.324e-02
##	2238	-0.2283908	-0.0375491	0.1283805	-1.336e-01
##	2338	-0.4125549	0.2751761	0.0618843	-2.497e-02
##	2437	0.0592927	0.0457217	0.3048821	6.599e-02
##	2535	0.0960123	0.2901334	-0.0455942	3.160e-01
##	2630	-0.2283908	-0.0375491	0.1283805	-1.336e-01
##	2730	0.0585523	0.2559429	-0.5520044	-3.666e-01
##	2830	0.0315217	-0.3283530	0.0196283	2.626e-01
##	2930	0.0215767	0.2516551	-0.6459980	1.376e-01
##	3029	-0.0320406	0.1260085	0.2754120	1.756e-01
##	3137	-0.4397088	-0.0126709	0.0948504	-2.155e-01
##	3229	-0.0155703	0.2202354	-0.0477499	3.147e-01
##	3329	-0.2044638	-0.0888181	0.1529247	2.172e-01
##	3429	-0.4058029	0.3607223	-0.2189515	-8.835e-02
##	3529	-0.0760729	-0.0497607	0.2603815	2.030e-01
##	3628	-0.1296553	-0.1323222	-0.3213732	2.393e-01
##	3728	-0.1855602	0.3483493	-0.1225912	5.071e-01
##	3829	-0.1177892	0.1138037	0.2968175	2.618e-01
##	3928	0.1189865	-0.3320652	0.1630039	2.934e-01
##	4029	-0.1176961	0.2414757	-0.1742623	5.960e-02
##	4138	-0.3905765	0.3775593	-0.1742943	-2.535e-01
##	4229	-0.0348588	0.1668644	-0.0288335	3.852e-02
##	4329	0.0969242	0.1741097	-0.1877912	4.181e-01
##	4429	-0.4096313	-0.1493904	0.1624654	-2.964e-01
##	4529	-0.0540087	-0.1590461	0.3119989	-3.275e-02
##	4628	-0.3431568	0.0034499	0.1983001	6.006e-02
##	4728	-0.2895133	-0.0052846	0.1653261	1.873e-02
##	4828	0.2706719	-0.2139981	-0.0199284	3.006e-01
##	4928	-0.5820999	-0.2871438	0.3480717	-5.189e-01
##	5028	0.0334025	0.0585847	0.4213796	2.334e-01
##	5136	-0.4679656	0.2462847	0.3459864	-7.362e-02
##	5228	-0.4520223	0.1214430	0.1840803	3.453e-02
##	5328	0.2675771	-0.0355162	0.0006194	-1.667e-01
##	5428	-0.4398386	0.2120736	0.0292504	-9.263e-02
##	5528	-0.6598907	0.6935585	-0.3041073	-3.721e-01
##	5628	0.1228072	0.2023907	-0.1132188	1.418e-01
##	5728	0.7165890	0.5893477	0.1003237	-6.878e-01
##	5828	0.2352144	-0.0139041	-0.3740529	5.572e-03
##	5928	0.2690503	0.3407191	0.0218824	-2.347e-01
##	6028	0.0349720	0.2594296	-0.0241657	2.486e-01
##	6136	-0.2566602	-0.1441207	0.2042839	-1.219e-01
##	6227	-0.2965964	-0.1610262	0.1945135	-4.480e-02
##	6327	0.1406624	0.2259286	0.1845741	-2.009e-01
##	6427	-0.2632741	0.1877905	0.1179856	1.111e-01
##	6526	0.6031663	-0.0068006	0.4351328	-1.506e-02
##	6626	-0.2444674	0.0796862	0.2191726	4.239e-02
##	6725	0.0382000	0.5432922	-0.1942198	-5.942e-01
##	6825	-0.3864826	-0.0102970	0.2835475	-2.647e-01
##	6925	-0.0481066	0.2263446	0.2380265	8.579e-02

##	7023	0.1526674	0.0841288	0.4048314	9.355e-02
##	7129	-0.2438400	-0.7813955	-0.0778077	-1.231e-01
##	7223	-0.3703753	-0.0062032	0.5345634	-8.262e-03
##	7323	-0.2332947	-0.6230070	0.1372089	-3.888e-01
##	7423	0.1297019	0.1550557	-0.1472036	5.770e-02
##	7523	-0.0798916	-0.1873270	0.2374266	2.435e-01
##	7623	-0.2726002	0.1246970	0.0713333	5.638e-02
##	7723	-0.1211563	-0.1365624	0.2374754	3.243e-02
##	7823	0.0613332	-0.2527733	0.2235868	1.054e-01
##	7923	-0.2458638	-0.0610742	0.0605757	1.108e-01
##	8023	-0.3046492	-0.0324591	0.1643438	-5.957e-02
##	8129	0.0444498	-0.0166096	0.5045293	-2.249e-02
##	8222	0.3979411	0.5157519	0.1340822	-2.920e-01
##	8323	-0.1615627	-0.1803350	0.2885832	2.732e-01
##	8423	-0.2835621	0.3843158	-0.2329181	-2.570e-01
##	8522	0.6491125	-0.0587328	0.2412793	-2.749e-01
##	8623	-0.1507930	0.0379517	-0.0144627	-2.957e-01
##	8722	0.1961413	-0.2455479	-0.0381623	4.445e-01
##	8823	-0.2895133	-0.0052846	0.1653261	1.873e-02
##	8923	0.0376807	0.1443428	0.0097291	4.941e-01
##	9023	-0.1964282	0.3965705	-0.3086188	-1.035e-01
##	9128	-0.3668176	-0.1493358	-0.0509618	-3.919e-01
##	9222	-0.0819419	-0.1061961	0.1050838	1.147e-02
##	9322	-0.2910824	-0.3012458	-0.1477966	6.761e-02
##	9420	0.0998779	0.1139785	-0.1739891	7.199e-02
##	9520	-0.0645343	-0.1563086	-0.3314003	-6.765e-03
##	9620	-0.1160040	0.2353798	-0.1229283	-2.031e-01
##	9722	0.1182470	0.4691932	0.1648001	1.334e-01
##	9822	0.0611955	0.0328720	-0.0062472	-1.131e-02
##	9922	-0.2695558	-0.3471605	-0.1425196	-2.266e-01
##	10022	0.1985055	-0.0366366	0.2577905	7.077e-02
##	10128	-0.1854305	0.1236048	-0.0569911	3.843e-01
##	10222	-0.1133914	-0.0075490	-0.2469025	4.332e-01
##	10322	-0.2423864	-0.0769265	-0.0068136	3.303e-01
##	10422	-0.2694828	-0.1950013	-0.5326334	8.340e-02
##	10522	0.1384899	0.0978424	0.0989479	3.634e-01
##	10622	-0.3455992	0.0211063	0.0014173	9.689e-02
##	10722	-0.1888588	-0.1918898	0.2819052	3.259e-01
##	10822	-0.1567642	0.0443601	0.0942504	1.964e-01
##	10920	0.1367451	0.2713506	-0.0631185	1.952e-01
##	11020	-0.1989917	0.2340113	0.1005444	2.229e-01
##	11125	-0.2836329	0.2888499	-0.0121579	1.153e-02
##	11220	-0.2138584	0.5579553	-0.2113633	2.500e-01
##	11319	-0.4416222	-0.1912018	0.2819913	-1.725e-01
##	11420	0.1300320	0.1450881	0.1754977	2.607e-01
##	11520	0.0701930	0.2618631	0.1622252	2.599e-01
##	11620	-0.2646073	0.0971431	0.0234155	1.224e-01
##	11720	0.1455881	-0.1473828	0.2094969	4.184e-01
##	11820	0.0425656	0.1618848	0.4142068	1.558e-01
##	11920	-0.0022415	0.0028141	0.3604585	3.493e-01
##	12020	-0.0792630	0.1442370	0.1756964	7.471e-02
##	12121	-0.2962662	0.2590603	0.2403905	-1.865e-02

##	12220	0.4873889	0.0964471	0.3196559	9.474e-02
##	12320	0.4254501	-0.1508024	0.1071930	8.383e-02
##	12419	0.0915167	0.2266679	-0.0450604	3.477e-02
##	12519	0.0008473	0.3310828	0.0561715	3.123e-01
##	12619	0.2761916	0.1967260	-0.1914867	4.057e-02
##	12719	0.0290240	0.2259687	0.1766327	1.450e-01
##	12819	-0.0326160	-0.1082948	0.1302640	3.967e-01
##	12917	0.1454341	-0.1309961	0.1387259	3.185e-01
##	13014	-0.3898341	-0.5518540	-0.1043521	-1.003e+00
##	13116	-0.0874457	0.0352279	0.1110728	6.491e-01
##	13214	0.2747454	-0.2686594	-0.1728071	1.650e-01
##	13312	-0.2074552	0.1263470	0.2287254	1.110e-01
##	13411	-0.1994243	0.0459706	0.0893016	3.445e-01
##	13511	-0.0464650	0.0831189	0.3041005	-1.098e-01
##	13610	-0.3574069	0.2370188	0.0102276	-1.218e-01
##	13710	-0.0779452	0.2298673	0.0789350	-3.552e-02
##	13810	-0.4561558	0.0796678	0.1551220	-1.600e-02
##	13910	-0.0894210	-0.2567719	0.0652943	1.640e-01
##	14010	-0.4085640	-0.1198440	0.4573837	-7.802e-02
##	14116	-0.2442445	0.1848813	0.2611788	3.364e-01
##	14213	-0.0835079	0.1584748	0.1789273	1.374e-01
##	14311	-0.1354962	0.1364674	0.1662084	2.377e-01
##	14410	0.3675657	0.2272349	-0.0033997	1.726e-02
##	14510	-0.3082028	-0.2247086	0.2587408	5.200e-02
##	14610	0.0579336	0.0396028	-0.0676779	3.039e-01
##	14710	-0.0414785	-0.0351164	0.2688450	1.363e-01
##	14810	-0.0811295	0.0929097	0.4110895	1.967e-01
##	14910	-0.0837463	-0.0671813	0.2026799	4.029e-01
##	15010	-0.1176830	0.0129151	-0.1601890	1.505e-01
##	15116	0.0189192	-0.3343325	-0.0675097	1.022e-01
##	15213	0.3056269	-0.0200145	-0.2158626	6.350e-03
##	15311	0.6833349	0.4702748	0.2080142	-6.914e-01
##	15410	-0.2510789	0.1701038	0.2578838	3.023e-01
##	1559	-0.3982615	0.0141365	-0.1989443	-9.418e-02
##	15610	0.3284256	-0.0780968	0.3148607	-2.435e-01
##	15710	-0.0234464	-0.0635550	0.2089241	2.241e-01
##	15810	-0.3745707	-0.3102277	0.2604356	-1.638e-01
##	1599	-0.2959559	-0.1116600	0.4071579	-1.281e-01
##	16010	-0.2540406	-0.3395770	0.1230849	-1.785e-02
##	16116	0.1984717	0.1927428	0.0342992	1.596e-01
##	16213	0.0876120	0.1391221	0.2123515	1.860e-01
##	16310	-0.0995648	0.0859285	-0.0007160	-2.793e-01
##	16410	-0.0908099	0.2061845	-0.0549381	4.914e-01
##	16510	0.3439367	0.0302291	0.3256173	1.625e-01
##	16610	-0.2946180	0.3501959	0.0872411	-1.218e-01
##	16710	-0.3656910	0.1572183	-0.0755572	-8.073e-02
##	16810	-0.2034983	0.0166446	0.2656824	1.933e-01
##	16910	0.2233807	-0.1225638	0.2891615	-4.074e-02
##	17010	-0.3437468	0.1141551	0.2665691	2.640e-01
##	17116	-0.0227390	0.0718352	0.0117870	3.198e-01
##	17213	0.1109895	-0.1605710	0.0293771	2.100e-01
##	17311	0.5602653	0.0847163	0.2994744	-7.107e-02

##	17410	-0.0486846	-0.4754710	-0.0518828	1.305e-01
##	17510	-0.4587410	-0.0128061	0.2987179	-1.435e-01
##	17610	-0.3105246	0.1821964	0.1076534	2.629e-01
##	17710	0.0405872	0.0306567	0.2094628	4.764e-01
##	17810	0.0429200	-0.2007502	0.0976714	-9.144e-02
##	17910	-0.4895070	0.0978410	0.4096915	-4.744e-02
##	1809	-0.2197105	0.2063950	0.3562036	-5.875e-02
##	18116	0.0604231	0.1971218	0.1263765	1.505e-01
##	18213	0.7822450	0.1442800	0.5868449	-4.987e-01
##	18311	0.1501269	-0.4002469	-0.7075978	-3.698e-02
##	18410	0.1697005	-0.3131264	-0.1985063	3.678e-01
##	18510	0.0538298	0.1590482	0.1073695	7.923e-02
##	18610	0.2584663	-0.4425188	-0.1309691	2.757e-01
##	1879	0.8556916	0.2470054	0.5605869	-4.774e-01
##	1888	-0.0432348	0.3463206	0.0943887	3.334e-01
##	1898	-0.1528427	0.2033218	0.0220443	3.056e-01
##	1908	-0.3860652	-0.1220834	-0.0659720	-1.461e-01
##	19115	-0.0288694	0.2325101	0.3712990	2.260e-01
##	19212	-0.0721898	0.2208979	0.0354161	1.023e-01
##	1939	0.2980011	0.1574224	0.2836581	1.089e-02
##	1948	0.5760435	0.2155334	0.1567522	-5.706e-01
##	1958	0.0390443	0.0284796	0.1062255	2.886e-01
##	1968	-0.0696477	0.2841943	-0.6047965	3.557e-01
##	1978	-0.2150904	-0.0653718	-0.0001356	2.777e-01
##	1988	-0.4919558	-0.3730378	0.3297822	-4.369e-01
##	1998	-0.2932670	-0.2658987	0.5711168	-8.663e-02
##	2008	-0.1714859	0.0320065	0.4576201	-4.045e-02
##	20115	-0.1003698	-0.0727738	0.3239166	4.386e-01
##	20212	0.2723503	0.0509715	-0.3210874	-3.710e-02
##	2039	-0.1962653	0.0759822	0.3120139	2.380e-01
##	2048	0.0721819	-0.2956380	-0.2894310	4.816e-01
##	2058	-0.1555740	0.0787998	0.3207413	9.882e-02
##	2068	-0.1018857	0.0117644	0.4004452	-7.047e-02
##	2078	-0.0985797	-0.0826581	0.6695857	1.712e-01
##	2088	-0.0844918	0.1855673	-0.0235546	2.386e-01
##	2098	0.2490261	0.4143857	-0.2166361	2.471e-02
##	21014	-0.2707723	-0.5430594	0.0186998	-4.377e-01
##	21115	0.4460202	0.3652148	0.1843507	-4.086e-01
##	21213	-0.3588144	0.0082276	0.2393964	-5.868e-02
##	21311	-0.2791404	0.0099118	-0.0446491	6.735e-02
##	21410	0.2905651	-0.0845543	0.0733102	-1.367e-01
##	2158	-0.1945577	0.1193582	-0.2910636	2.359e-01
##	2168	-0.0471089	0.0970858	0.0739329	2.250e-01
##	2178	-0.2244250	0.2229476	0.2011948	-2.083e-01
##	2188	-0.2237895	-0.2890216	0.1294390	1.017e-01
##	2198	-0.3455992	0.0211063	0.0014173	9.689e-02
##	2208	0.2223845	-0.0935456	-0.0339463	-1.527e-01
##	22115	-0.2599126	0.0670307	0.3673098	1.259e-01
##	22212	0.2214902	-0.0007283	0.1924310	-9.381e-02
##	2239	-0.5527446	0.1802766	0.3423679	-1.250e-01
##	2248	-0.0595980	0.0051982	-0.1588129	-5.245e-02
##	2258	-0.2770286	-0.2905388	0.0503941	-4.786e-02

##	2268	0.4630467	0.2352317	0.1783506	-2.631e-01
##	2278	-0.0087946	0.1145822	0.1973306	4.084e-01
##	2288	0.0554407	0.2496381	0.3584665	1.279e-01
##	2298	-0.3388942	0.2399566	-0.0672323	1.874e-01
##	2308	-0.2132582	0.0771280	0.0337945	-1.586e-01
##	23115	-0.1962302	-0.3151050	0.1372282	-1.294e-01
##	23212	-0.4039433	-0.0290941	0.2343488	-8.205e-02
##	2339	-0.1768456	-0.0202879	0.2212933	1.487e-01
##	2348	-0.1174084	-0.0192737	-0.1729562	5.914e-02
##	2358	0.6213048	0.0990737	0.1092882	-3.735e-01
##	2368	-0.0837858	-0.0285005	-0.0417168	2.686e-01
##	2378	-0.3028156	0.0607948	0.0123554	1.112e-01
##	2388	-0.5407475	-0.3096353	-0.0410472	-6.891e-01
##	2398	-0.2646073	0.0971431	0.0234155	1.224e-01
##	2408	0.0326716	0.2739125	-0.2604332	1.770e-01
##	24114	0.5552925	-0.0935145	0.0497040	-2.488e-01
##	24211	-0.2267318	-0.1619887	-0.0176708	-5.496e-02
##	2438	-0.2616674	0.0334369	0.0231556	-1.770e-01
##	2447	-0.5174936	-0.0322362	0.1284530	-2.398e-01
##	2457	0.4542766	0.1067544	0.2419394	8.968e-02
##	2467	-0.1796916	0.2280367	0.1020720	1.350e-01
##	2477	-0.1796916	0.2280367	0.1020720	1.350e-01
##	2487	-0.0243008	-0.2554836	0.2363398	-2.189e-01
##	2497	0.1708117	-0.2555332	-0.4081439	1.613e-01
##	2507	-0.2691186	0.0159881	0.2201018	-1.857e-01
##	25114	0.3845035	0.4539931	0.3024619	-3.594e-01
##	25211	-0.2526786	0.0678462	-0.0312029	1.995e-01
##	2536	-0.3306898	0.1370951	-0.1200978	-7.072e-02
##	2545	-0.4792284	-0.0676265	0.0369557	-9.856e-02
##	2555	0.1830823	0.2318946	0.1477203	-1.262e-01
##	2565	0.3078647	0.1837134	0.1962097	-1.487e-01
##	2575	-0.0270076	0.1429608	0.0963218	2.794e-01
##	2585	-0.1176830	0.0129151	-0.1601890	1.505e-01
##	2595	0.0431937	0.1365896	0.2100495	3.081e-01
##	2605	-0.0754671	-0.0650485	-0.2832226	2.266e-01
##	26112	0.0661842	0.0684677	-0.2029694	8.651e-02
##	26210	-0.3183032	0.0326611	0.0080953	4.420e-02
##	2635	-0.1305890	0.2589601	0.2670902	3.343e-01
##	2645	-0.2236141	0.1513053	0.0394272	1.854e-01
##	2655	-0.0385683	-0.1788363	-0.2020051	1.711e-01
##	2665	-0.1181587	0.3627796	-0.1142527	3.488e-01
##	2675	-0.2005555	0.4834639	-0.2046020	-1.592e-01
##	2685	-0.0384989	0.0790561	0.3456930	3.077e-01
##	2695	-0.4198811	-0.1298023	-0.2785953	-3.380e-01
##	2705	-0.3183032	0.0326611	0.0080953	4.420e-02
##	27112	0.2105925	-0.5292327	-0.5163980	4.197e-02
##	27210	-0.3016037	0.0165313	0.0133852	-9.993e-02
##	2735	-0.3335410	-0.2203218	0.0542631	-3.484e-01
##	2745	-0.2724793	0.0774060	-0.2431008	4.479e-02
##	2755	0.1962104	0.5177002	-0.0870639	6.504e-02
##	2765	-0.0677303	0.3859928	-0.4614034	1.752e-01
##	2775	0.0954911	0.1688148	0.0238724	3.825e-01

##	2785	-0.1723844	-0.1845392	0.2237287	1.607e-02
##	2795	-0.1963912	0.4212164	-0.2115918	-4.252e-01
##	2805	0.0275359	0.0174529	-0.2345136	1.010e-01
##	28112	0.0337189	0.0636917	-0.2779383	4.907e-01
##	28210	0.2071115	0.0934696	-0.1905805	-2.456e-01
##	2835	-0.2164806	0.2860605	-0.1186874	2.896e-01
##	2844	0.1208897	0.1005922	-0.2566120	3.224e-01
##	2855	0.0005289	0.3037009	-0.2363739	2.952e-01
##	2865	-0.4427579	-0.1119520	-0.1757352	-6.534e-01
##	2875	0.0618482	-0.1415298	0.3934792	2.940e-01
##	2885	0.0486157	0.3181848	-0.1807189	-2.124e-02
##	2895	0.1135257	0.2840026	-0.2579722	-1.225e-02
##	2905	0.0872684	0.0331459	-0.5354105	1.523e-01
##	29112	-0.2359940	0.2137462	0.1258873	1.190e-01
##	29210	-0.1456052	0.2994976	-0.1077076	1.052e-01
##	2935	0.3456839	0.4167805	-0.0043066	1.896e-02
##	2945	-0.1143339	-0.4097315	0.0418453	-3.639e-01
##	2955	0.2332906	0.2882429	-0.3271451	-1.544e-02
##	2965	-0.3292278	0.3766994	0.1051153	1.592e-02
##	2975	-0.1037242	0.2743393	0.1238516	3.895e-02
##	2985	0.3473659	-0.9010746	0.2704460	5.619e-01
##	2995	-0.4165200	-0.2505045	0.2568108	-4.016e-01
##	3004	0.1624503	0.5129521	-0.0577336	5.580e-03
##	30112	0.1834036	-0.4712329	-0.6023729	6.467e-03
##	30210	0.3650183	-0.1602142	0.2858128	-5.919e-01
##	3034	0.0526066	0.1477673	-0.4284718	2.956e-01
##	3044	-0.2251774	-0.0318647	-0.3469471	-1.228e-01
##	3054	0.3964331	0.4342047	-0.2185837	-2.850e-01
##	3064	0.0648297	0.0596833	-0.3989552	-2.750e-01
##	3074	0.0820423	0.1226693	-0.4547982	2.166e-01
##	3083	0.2398802	0.0144618	0.1415768	1.527e-01
##	3094	-0.0486846	-0.4754710	-0.0518828	1.305e-01
##	31011	-0.2263986	0.4577103	-0.2137522	-1.647e-01
##	31111	-0.1354072	0.3465111	-0.0514462	4.061e-02
##	31211	0.1447168	0.2428811	0.0150996	-3.081e-01
##	3138	-0.0672018	-0.1639667	0.1006813	-2.749e-01
##	3144	0.0753460	0.5607101	0.0291417	7.744e-02
##	3154	-0.0496081	0.2966495	-0.1928156	2.069e-01
##	3164	-0.0402153	0.0152164	-0.4246545	-7.221e-02
##	3174	0.1282891	0.1076966	0.1522766	1.980e-01
##	3184	0.0949785	0.1232783	-0.0280396	-5.257e-01
##	3194	-0.0169709	-0.0966348	0.3280325	7.904e-02
##	3204	0.3887876	0.2330818	0.2946394	5.685e-02
##	32111	0.1737207	0.2756383	0.0308751	-3.090e-01
##	32210	-0.2144774	0.3803476	0.0445919	-1.482e-02
##	3234	0.6959961	0.3187550	0.1759133	-5.158e-01
##	3244	0.0946341	-0.0930303	-0.6490272	6.179e-02
##	3254	-0.1069000	0.1144689	-0.0549666	1.152e-01
##	3264	-0.1428164	0.5125072	-0.2248716	-9.545e-02
##	3284	-0.3547428	0.1686650	-0.1237989	1.576e-01
##	3294	0.0739180	-0.6967663	-0.4246726	4.446e-01
##	3304	-0.2325431	0.0943115	-0.2333304	-3.230e-02

33111 -0.0137296 -0.2814874 -0.2478170 -1.638e-01
33210 -0.1315524 -0.3032239 -0.0294444 -1.524e-01
3334 -0.4697931 0.0233553 0.3178733 -3.267e-02
3344 0.3503873 -0.1795165 -0.3642108 -1.854e-01
3353 -0.2316527 -0.0308183 0.0669498 1.817e-01
3364 0.4480666 -0.0088733 -0.2248949 -2.976e-01
3374 -0.3055540 -0.1730744 0.0508088 -1.938e-01
3384 0.4480666 -0.0088733 -0.2248949 -2.976e-01
3394 -0.3289936 0.2163742 -0.1115166 2.500e-01
3404 0.2962023 0.3710381 0.2255137 -1.502e-01
34111 0.1140350 0.0790020 0.4727919 3.654e-01
34210 -0.2080355 0.0847699 0.3108540 3.162e-01
3434 0.0985469 0.0468556 0.1717029 -8.574e-02
3444 -0.6505531 -0.4153053 -0.1849085 -1.428e+00
3454 0.2833557 -0.5555769 0.1608860 5.162e-01
3464 -0.0547823 0.0796652 0.0162313 4.249e-01
3474 0.0478275 -0.0716038 0.0567101 2.662e-02
3484 -0.1489997 -0.1943676 -0.1060638 -2.339e-02
3494 -0.0274863 0.0912200 0.0229093 3.722e-01
3504 0.7866216 0.5291781 0.3796103 -7.143e-01
35111 -0.3713799 0.3963378 -0.2492218 1.152e-01
35210 0.4002430 0.1209370 0.4348255 3.738e-01
3534 -0.3168094 -0.0168394 0.1586481 7.142e-02
3544 0.1619122 -0.4627639 0.5861206 3.627e-01
3554 -0.1227538 0.0508920 -0.0003979 5.561e-01
3564 0.0416503 -0.0269902 -0.1118878 2.416e-01
3574 -0.2240882 -0.5784204 0.5157260 -3.679e-01
3584 0.0571648 -0.5308488 0.2693757 4.832e-01
3594 0.0595485 0.2269247 -0.2288966 -2.546e-01
3604 -0.1449056 0.3217816 -0.4422510 8.921e-03
36111 -0.0532534 0.0589180 0.0043217 3.788e-01
3629 -0.2111123 0.1286709 -0.5199130 -3.457e-01
3634 0.2376701 0.1192441 0.2191120 2.924e-01
3644 -0.1718765 -0.1765173 -0.0032037 -3.388e-01
3654 -0.3525558 0.0885115 -0.1877729 7.024e-02
3664 0.3153747 0.2925978 -0.3341940 -1.918e-01
3674 0.0806131 0.5118482 -0.3784209 -3.094e-02
3684 0.1325694 -0.1344847 -0.3091088 -5.539e-01
3694 -0.2820252 0.1024092 0.0689403 -9.032e-03
3704 -0.2075440 0.0522482 -0.3201705 9.431e-02
37111 0.4199068 0.0189105 0.4989097 -3.776e-02
3729 -0.2996862 0.1183297 0.1567783 -2.805e-01
3734 0.3258956 -0.7751984 -0.1517950 1.925e-01
3744 -0.5041294 0.1505278 0.2583668 -2.012e-02
3754 0.1016076 0.3260361 0.2009387 2.938e-01
3764 -0.1433220 0.0500812 -0.0796612 3.143e-02
3774 0.3279351 -0.3821242 -0.0046146 2.344e-01
3784 0.2478062 -0.2296037 -0.5682819 -1.492e-01
3794 -0.3999590 -0.1163129 0.3227682 -1.272e-01
3804 -0.0472801 -0.2208176 -0.0018549 -1.318e-01
38111 0.3958550 -0.4107764 0.2940348 -3.267e-01
38210 -0.1092189 -0.4680670 -0.2143178 -1.427e-01

##	3834	-0.0072863	0.3000359	0.0380349	2.736e-01
##	3844	-0.1132727	0.3599018	-0.4216253	3.632e-03
##	3854	0.0631513	-0.2052863	-0.0977962	6.275e-02
##	3864	0.3311375	-0.3429839	0.1712715	3.796e-01
##	3874	-0.2028843	0.1422178	0.0411575	7.841e-02
##	3884	0.7277959	0.6089260	0.1163078	-7.007e-01
##	3894	0.1512771	0.0586585	0.2249674	2.280e-01
##	3904	0.1087387	-0.0927304	-0.1131694	5.217e-01
##	39111	-0.2201091	-0.0987430	-0.2101523	1.463e-01
##	3929	-0.0480681	0.1610284	-0.3353467	2.188e-01
##	3934	0.3723937	0.4305260	0.1141812	-5.231e-01
##	3944	0.4043500	-0.9425581	0.0962878	2.949e-01
##	3954	0.2352144	-0.0139041	-0.3740529	5.572e-03
##	3964	0.2646981	0.0576197	0.0622302	-7.231e-03
##	3974	0.0867061	-0.2900974	0.3808868	3.619e-01
##	3984	-0.1696845	-0.3390812	-0.4055177	-2.216e-01
##	3994	-0.0595980	0.0051982	-0.1588129	-5.245e-02
##	4004	-0.0405335	-0.0227263	-0.1662192	-2.004e-01
##	40111	0.0645058	-0.1965019	0.0981897	4.243e-01
##	40210	0.0958293	0.0088355	0.2466130	5.630e-02
##	4034	0.1375351	-0.1845472	-0.5133688	1.178e-01
##	4044	-0.1605841	0.1044319	-0.2958474	9.444e-02
##	4054	0.4171997	0.3292972	0.2872537	-1.920e-01
##	4064	0.4234134	-0.9832242	0.1565685	3.198e-01
##	4074	0.1072259	-0.3087300	-0.8432562	-9.299e-02
##	4084	-0.1641278	-0.2162884	0.0062211	-5.852e-01
##	4094	-0.2325431	0.0943115	-0.2333304	-3.230e-02
##	41011	-0.3539274	-0.7081619	0.5497017	-4.452e-01
##	41111	-0.1049641	0.3639996	-0.0783938	1.571e-01
##	41211	0.3749295	-0.6018247	0.5816195	1.299e-01
##	4139	0.2105925	-0.5292327	-0.5163980	4.197e-02
##	4143	-0.1133495	0.1244270	-0.2842915	3.263e-03
##	4154	0.4858009	0.0915154	0.0760967	-5.278e-01
##	4163	0.3986627	-0.5514123	-0.0912546	4.775e-02
##	4174	-0.0372054	0.2127908	-0.2394952	-3.315e-03
##	4184	0.3551241	-0.4401240	0.0813604	2.699e-01
##	4194	0.3133038	-0.5594987	0.0424340	3.473e-01
##	4204	0.2034312	-0.0968089	0.1490109	-2.901e-01
##	42111	0.0818038	-0.1623871	0.4452064	-8.894e-02
##	42210	0.1980007	-0.3135330	-0.3221689	1.968e-01
##	4234	0.2879462	0.6223441	0.0388502	-2.736e-01
##	4244	-0.3970395	-0.3650900	-0.0317636	-4.511e-01
##	4254	-0.0724637	0.2105817	0.3866031	1.778e-01
##	4264	-0.1223744	-0.1387612	0.5945029	-8.243e-02
##	4274	0.0498737	0.0664716	-0.5006789	2.535e-01
##	4284	-0.3396572	-0.6588919	0.0543136	-9.377e-01
##	4294	-0.3500860	0.0541466	0.0534232	2.798e-02
##	4304	0.5990314	0.1664953	0.2294430	-5.406e-01
##	43111	0.0756564	-1.0038593	0.4085728	-2.870e-01
##	43210	-0.6109245	0.1613132	0.1397418	-3.456e-01
##	4334	0.1483909	0.0008814	-0.5723561	2.692e-05
##	4344	0.5711279	0.2061116	-0.0493776	-4.390e-01

## 4354	0.2621266	0.0361904	-0.0185207	2.634e-01
## 4364	0.2054331	-0.1715000	0.1105786	9.211e-02
## 4374	0.1897033	0.1214997	-0.0170889	8.721e-02
## 4384	0.7788826	-0.0666454	0.5313253	-2.756e-01
## 4394	0.8291214	0.2557382	0.4980186	-4.817e-01
## 4404	0.1267685	0.3069327	-0.1864923	-2.641e-01
## 44111	0.3377955	0.0361832	-0.1699818	-3.061e-02
## 44210	0.0452857	0.3903285	-0.1359353	1.185e-01
## 4434	0.4215284	-0.5358067	0.4570989	4.976e-01
## 4444	0.6794530	0.5244722	0.0473583	-6.452e-01
## 4454	0.4346263	-0.6281303	0.0224670	9.470e-02
## 4464	-0.3297889	0.2051361	0.1729002	1.293e-01
## 4474	-0.1232253	0.0967379	0.1138392	2.559e-01
## 4484	-0.4096313	-0.1493904	0.1624654	-2.964e-01
## 4494	-0.0372054	0.2127908	-0.2394952	-3.315e-03
## 4504	-0.0628108	-0.0009098	0.0371195	-1.631e-02
## 45111	-0.0564503	-0.0556265	0.2093579	3.503e-01
## 45210	0.2611253	0.0412862	-0.1552113	1.396e-01
## 4534	-0.1102572	-0.6853238	0.8605988	-3.340e-01
## 4544	-0.4289662	-0.0305665	-0.2016304	-2.021e-01
## 4554	-0.0564503	-0.0556265	0.2093579	3.503e-01
## 4564	0.5238510	-0.3212425	0.6383113	-4.529e-01
## 4574	0.0433459	-0.0435856	-0.5980553	2.028e-01
## 4584	0.2294493	0.1424883	-0.4567458	-9.311e-02
## 4594	-0.3682156	0.1794034	0.2342566	-1.570e-01
## 4604	0.3452720	0.0149795	0.1745367	2.732e-01
## 46111	0.6529509	0.1752112	0.4570418	-6.191e-01
## 4629	-0.2418228	0.2532747	-0.0901903	8.177e-02
## 4634	-0.0536964	-0.2007572	-0.0897202	-3.906e-01
## 4644	0.2352595	-0.1234749	0.2956260	3.338e-01
## 4654	-0.2310494	0.3392870	0.0036540	2.616e-02
## 4664	0.5007959	0.1846839	0.4313823	-3.809e-02
## 4674	0.0322798	-0.6527124	0.4886137	6.154e-02
## 4684	-0.1033222	0.5851439	-0.2218813	1.149e-01
## 4694	-0.1519190	-0.5183932	-0.3110494	-5.842e-01
## 4704	0.3328936	0.1567391	-0.2347369	-1.067e-01
## 47111	-0.1183681	0.0264684	-0.4188810	1.706e-01
## 4729	0.1912919	-0.0906355	-0.4366977	5.604e-02
## 4734	-0.1098647	0.2228913	-0.1808546	-1.792e-02
## 4744	-0.2104108	0.0999110	0.0349622	-6.714e-02
## 4753	0.1132577	-0.4835672	-0.0347263	-4.333e-01
## 4764	0.2956800	-0.1428898	-0.1828530	8.452e-02
## 4774	0.1024226	0.1769085	-0.0129988	5.656e-01
## 4784	0.2534935	-0.6207495	-0.3807395	9.798e-02
## 4794	0.0862469	-0.1351024	-0.4623969	2.588e-01
## 4804	-0.4730606	0.5420350	-0.0153750	1.853e-02
## 48111	0.1195235	-0.2060884	-0.3571721	3.023e-01
## 4829	-0.0704485	0.0329101	-0.1486330	5.928e-02
## 4834	-0.0817875	-0.1252621	-0.0792282	-1.308e-02
## 4844	-0.1494678	0.1669408	0.3656041	2.854e-01
## 4854	0.0880015	-0.1115329	-0.0433137	1.800e-01
## 4864	-0.1921059	0.0084149	0.0806999	1.903e-01

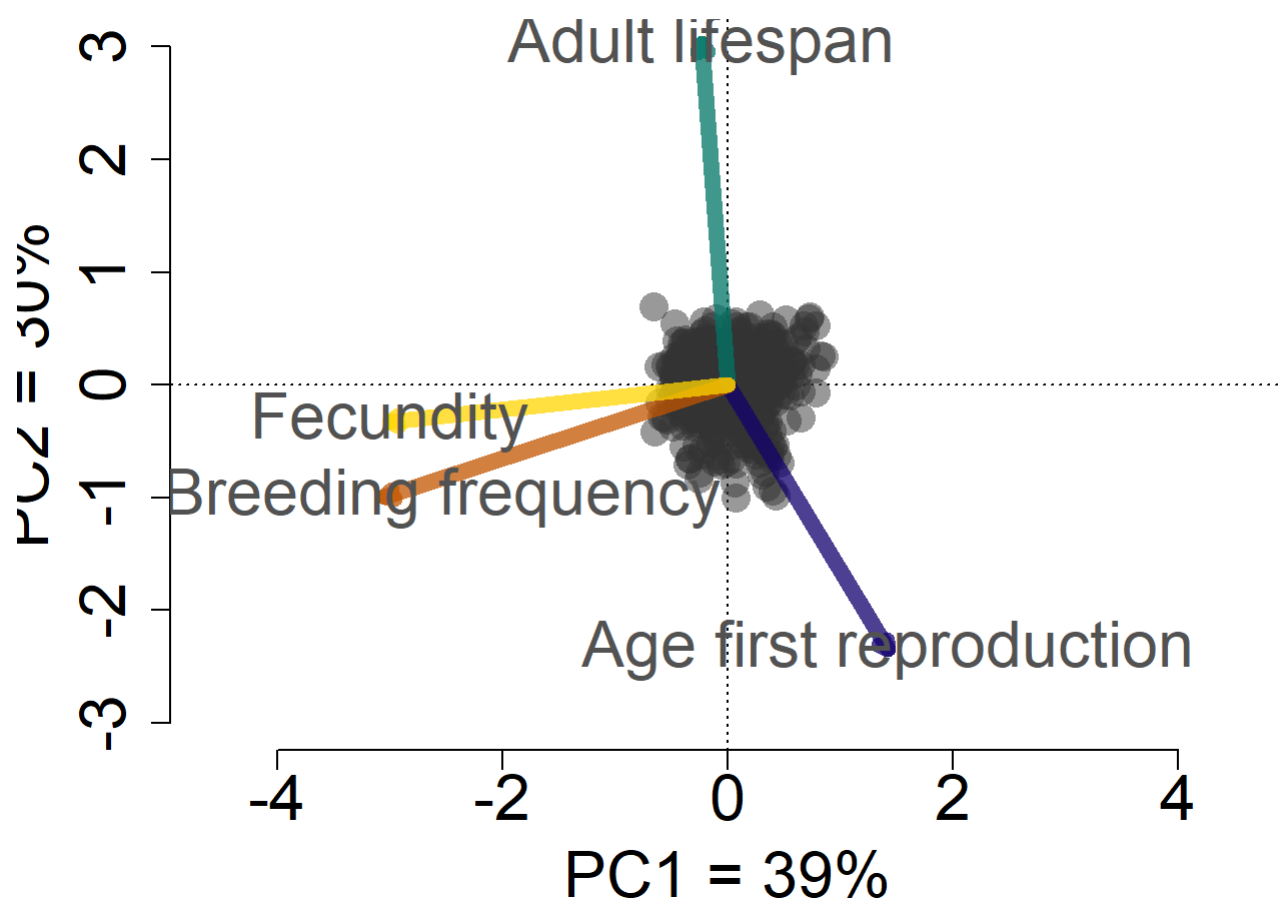
## 4874	0.1753575	0.3357122	0.0210495	6.246e-03
## 4884	0.4269233	0.2254328	0.1284561	1.580e-02
## 4894	-0.3920910	0.0792136	0.1341715	-1.021e-01
## 4904	0.3240727	-0.5790179	0.5940710	3.200e-01
## 49111	-0.0116975	-0.2747788	0.1424052	7.843e-02
## 4929	-0.0744691	-0.1628178	-0.0878299	-1.673e-01
## 4934	-0.2391316	0.0611575	-0.2401350	-1.357e-01
## 4944	-0.1962302	-0.3151050	0.1372282	-1.294e-01
## 4954	0.2043432	-0.0774317	0.0072786	4.314e-01
## 4964	-0.3830504	0.0834215	-0.4535207	-3.734e-01
## 4974	0.2478062	-0.2296037	-0.5682819	-1.492e-01
## 4984	0.0373284	0.1593942	-0.0832111	4.741e-02
## 4994	-0.0561044	0.3539499	-0.0794514	-2.000e-02
## 5004	-0.0640971	0.2068400	0.0312211	3.412e-02
## 50111	-0.2418228	0.2532747	-0.0901903	8.177e-02
## 5029	-0.1282433	0.4041058	-0.0297060	1.877e-01
## 5034	-0.3306661	-0.2180890	0.3167752	-1.153e-01
## 5044	-0.5599590	-0.2959164	0.4508604	-4.366e-01
## 5054	0.0486037	0.2485717	0.1879818	1.435e-01
## 5064	0.3596801	0.4557343	-0.1485077	-3.981e-01
## 5074	-0.1811563	-0.0175541	0.1399364	-2.248e-01
## 5084	0.4056935	0.0492304	0.4539656	-5.631e-02
## 5094	-0.0949258	0.1386819	0.4752493	1.576e-01
## 51011	-0.0853423	0.3125889	0.1059524	2.437e-01
## 51111	0.2867702	-0.6917355	-0.2755147	1.414e-01
## 51211	-0.0660499	0.2079858	-0.3019053	3.344e-01
## 5137	0.1308222	-0.5210361	0.0208151	-4.104e-01
## 5144	0.2909071	0.3892274	0.0149231	-4.156e-02
## 5154	0.2478062	-0.2296037	-0.5682819	-1.492e-01
## 5164	0.0084807	0.2395356	-0.2836714	1.905e-01
## 5174	0.3346439	-0.6050217	0.1099142	3.752e-01
## 5184	0.3725792	-0.6464760	0.4498326	-2.405e-01
## 5194	-0.0168569	-0.6919633	-0.9457599	1.549e-01
## 5204	0.3795582	-0.0237767	0.1353716	1.755e-01
## 52111	-0.2592785	-0.8303565	0.3084811	-8.328e-01
## 5229	-0.1566332	-0.3258774	0.0384587	1.537e-01
## 5234	0.2352144	-0.0139041	-0.3740529	5.572e-03
## 5244	0.0661842	0.0684677	-0.2029694	8.651e-02
## 5254	0.2236635	0.3955201	-0.0070280	2.724e-01
## 5264	-0.1074353	-0.0647917	-0.4670587	-6.278e-02
## 5274	-0.0371719	-0.0380759	-0.0434082	1.027e-01
## 5284	0.3153747	0.2925978	-0.3341940	-1.918e-01
## 5294	0.6490347	-0.2934863	0.3461286	-1.264e-01
## 5304	-0.3469339	0.3201246	0.0432249	8.273e-02
## 53111	0.0271464	0.2681078	0.0211517	1.071e-01
## 5329	0.0482092	-0.0342380	-0.0353541	-1.529e-01
## 5334	0.4856431	0.1141605	0.2935038	-1.945e-01
## 5344	0.2761916	0.1967260	-0.1914867	4.057e-02
## 5354	-0.0316106	-0.2323000	0.0794376	5.243e-02
## 5364	0.4949503	0.3686145	-0.2902609	-5.385e-01
## 5373	0.4090249	0.2185051	-0.4128127	-4.398e-01
## 5384	0.4519259	0.1269882	-0.2771543	-3.837e-01

## 5394	0.0232831	0.1599846	-0.3386278	3.050e-02
## 5404	0.0524978	0.0897451	0.4671253	1.184e-01
## 54111	-0.2043644	0.3368129	0.0084628	-8.085e-02
## 5429	-0.2898585	-0.3701804	0.3905006	-2.061e-01
## 5434	0.3606851	0.1121723	0.2048370	-1.773e-02
## 5444	0.1634460	-0.1293570	-0.2945272	2.518e-01
## 5453	0.0431677	-0.3340505	-0.2563597	-5.248e-01
## 5464	0.2335805	-0.5782708	-0.4437072	7.198e-02
## 5474	0.5076183	-0.0884852	0.2357523	-4.344e-01
## 5484	0.2563221	0.1031255	0.2506666	2.523e-02
## 5494	0.2956800	-0.1428898	-0.1828530	8.452e-02
## 5504	0.1181853	0.0682882	0.0784427	3.760e-01
## 55111	-0.1920180	-0.1742381	-0.2206443	-2.313e-01
## 5529	0.1708117	-0.2555332	-0.4081439	1.613e-01
## 5534	0.3186679	-0.1919280	-0.1101622	1.145e-01
## 5544	0.4635801	-0.6898948	0.1140226	1.325e-01
## 5554	0.5661647	0.1928474	0.3022521	-6.546e-01
## 5564	0.2539964	0.1436592	0.2365303	2.811e-01
## 5574	0.1390923	-0.2679447	-0.1540953	4.612e-01
## 5584	0.0179973	0.0226973	-0.0774483	3.810e-01
## 5594	0.4005198	0.0128427	0.1652683	1.514e-01
## 5604	0.1388617	0.0734201	0.4378158	-6.507e-02
## 56111	-0.6044224	-0.1557565	-0.1182684	-6.630e-01
## 5629	-0.0911911	-0.1011445	0.0357947	3.708e-01
## 5634	0.1891618	-0.5635920	-0.2298153	3.553e-01
## 5644	-0.0539775	0.1984844	-0.0160893	1.797e-01
## 5654	0.1301694	-0.0583710	-0.3997521	2.084e-01
## 5664	-0.0963787	-0.2033082	-0.2161484	2.827e-01
## 5674	0.2723503	0.0509715	-0.3210874	-3.710e-02
## 5684	-0.3225848	-0.0455643	-0.2623209	-2.944e-01
## 5694	-0.2229358	-0.1962457	0.0653970	1.675e-01
## 5704	-0.3396572	-0.6588919	0.0543136	-9.377e-01
## 57111	0.4752556	-0.0668731	-0.1389199	-2.621e-01
## 5729	-0.3048626	0.1126474	0.3133516	-6.058e-02
## 5734	-0.0841687	-0.1149056	-0.1647583	4.243e-01
## 5744	0.1391958	0.1305979	0.4572975	1.092e-01
## 5754	0.2517575	-0.2196213	-0.2454979	1.350e-01
## 5764	0.0690731	0.0096882	0.1490319	5.718e-01
## 5774	0.0505595	0.0104133	0.0964328	2.175e-01
## 5784	0.3583519	-0.0403192	0.5217066	1.231e-01
## 5794	0.2956800	-0.1428898	-0.1828530	8.452e-02
## 5804	-0.4233815	0.1034909	0.2023299	-2.092e-01
## 58111	0.6010566	0.3875157	-0.0644553	-5.551e-01
## 5829	-0.0517551	0.1021738	0.1725734	-1.808e-01
## 5834	-0.3437034	-0.2923052	-0.0377120	-2.385e-01
## 5844	-0.0152793	0.0936833	-0.1826732	3.376e-01
## 5854	0.3351801	0.1308970	0.1660651	-3.319e-01
## 5864	0.2245685	-0.1616215	-0.3314728	9.949e-02
## 5874	0.0862469	-0.1351024	-0.4623969	2.588e-01
## 5884	-0.5172885	-0.1435018	-0.1939691	-5.095e-01
## 5894	0.0498737	0.0664716	-0.5006789	2.535e-01
## 5904	0.2136229	-0.2363949	-0.1358615	3.173e-01

##	59111	0.1717971	0.1842248	0.1621899	3.840e-01
##	5929	0.0911654	0.4732333	-0.1127631	2.678e-01
##	5934	-0.1489600	-0.2531115	0.4070397	-1.341e-01
##	5944	0.3708675	-0.0146188	-0.3927646	-2.906e-01
##	5954	0.3056269	-0.0200145	-0.2158626	6.350e-03
##	5964	0.2742097	-0.0170136	-0.6050941	-2.849e-01
##	5974	0.2327506	-0.0082838	-0.1956810	1.722e-01
##	5984	-0.4246281	-0.1738077	-0.0651051	-2.803e-01
##	5994	0.2679601	0.0508890	0.1236609	-3.225e-01
##	6004	-0.2073411	-0.0830605	0.2722818	-2.833e-01
##	60111	-0.0262575	0.0047056	-0.3226032	5.866e-01
##	6029	-0.1057883	-0.2588444	-0.2444093	1.806e-01
##	6034	0.4852025	0.0560022	-0.1719294	-3.403e-01
##	6044	0.0431677	-0.3340505	-0.2563597	-5.248e-01
##	6054	-0.2450051	0.0852667	0.0264987	-3.577e-04
##	6064	0.2810828	-0.3005897	-0.4630570	-1.058e-01
##	6074	0.3113022	-0.0657518	0.0034545	2.051e-01
##	6084	-0.1017421	-0.6254311	-0.1523837	-5.186e-01
##	6094	-0.0818290	-0.6679099	-0.0894160	-4.927e-01
##	61011	0.1501269	-0.4002469	-0.7075978	-3.698e-02
##	61111	-0.1679009	0.0641941	-0.6582586	-1.417e-01
##	61211	-0.1067906	0.1171792	-0.2077578	-3.912e-01
##	6137	-0.1519190	-0.5183932	-0.3110494	-5.842e-01
##	6144	-0.3457334	-0.1846576	-0.3834029	-3.261e-01
##	6154	-0.4266021	0.1783448	0.0943223	-5.758e-01
##	6164	-0.1332582	0.2998888	-0.4524736	-7.812e-02
##	6174	0.3567295	-0.3273136	0.1703151	-3.778e-01
##	6184	0.0682306	-0.3056204	-0.6122150	1.974e-01
##	6194	-0.1183681	0.0264684	-0.4188810	1.706e-01
##	6204	-0.1147459	-0.0163334	-0.4428883	7.166e-02
##	62111	0.2049052	-0.1380868	-0.7039403	-2.052e-01
##	6228	-0.0968785	0.2900013	-0.1517477	1.236e-01
##	6234	-0.3236825	0.2667367	0.0166110	-1.077e-01
##	6244	-0.1216981	0.0986121	-0.3740974	3.104e-01
##	6254	-0.3171258	-0.0151349	0.0130841	-2.008e-01
##	6264	0.3275406	-0.0802665	0.1673038	-6.409e-01
##	6274	-0.0718449	-0.1078503	-0.3072299	1.277e-01
##	6284	-0.0642644	-0.7053788	-0.0338746	-4.697e-01
##	6294	0.3005912	-0.8012945	0.1225387	5.008e-01
##	6304	-0.1067906	0.1171792	-0.2077578	-3.912e-01
##	63111	0.0946341	-0.0930303	-0.6490272	6.179e-02
##	6328	-0.5181274	0.1667540	0.1011323	-4.053e-01
##	6334	0.2245685	-0.1616215	-0.3314728	9.949e-02
##	6344	0.3279665	0.0768981	-0.5284230	-3.466e-01
##	6354	-0.2966657	0.3370384	0.0881538	-1.806e-01
##	6364	-0.4233815	0.1034909	0.2023299	-2.092e-01
##	6374	0.0983862	-0.1412759	0.1233116	-8.737e-02
##	6384	-0.2636608	-0.0038197	0.3376908	-5.620e-02
##	6394	0.3741619	-0.1013576	0.1581649	2.248e-01
##	6404	0.4147900	0.0621127	-0.3301198	-3.411e-01
##	64111	-0.0569072	-0.1135403	0.3182621	1.561e-01
##	6428	0.1994739	0.0627022	-0.3009059	1.287e-01

## 6434	0.0467603	-0.1797442	-1.0344560	-1.719e-01
## 6444	-0.4223105	0.1578807	0.0076088	-2.931e-01
## 6454	-0.3916909	0.3091036	0.0634923	6.562e-02
## 6464	-0.1968753	-0.2123303	-0.3637775	2.487e-01
## 6474	0.2790364	0.0734984	-0.0538114	-2.167e-01
## 6484	-0.1200271	0.1509080	-0.2728297	9.201e-02
## 6494	-0.1661410	0.2283284	-0.0793700	-2.914e-01
## 6504	-0.1364079	-0.4100672	-0.3002928	-1.782e-01
## 65111	-0.0044846	0.1088220	0.1315056	-9.753e-02
## 6527	-0.0048554	0.3066487	0.1104501	-2.727e-02
## 6534	0.2042325	0.1866055	-0.1289697	-8.618e-02
## 6543	0.1161707	-0.3697532	0.3969899	1.770e-01
## 6554	-0.4510564	0.2168680	0.0640956	-1.956e-01
## 6564	-0.1573753	0.0614566	-0.0148593	-1.677e-01
## 6574	0.3844494	-0.5210924	-0.1361988	2.919e-02
## 6584	-0.4157106	0.0189615	0.0927215	-1.274e-01
## 6594	-0.0043791	-0.3123345	0.1338035	-7.575e-02
## 6604	0.2473992	0.3126992	0.0050099	-2.276e-01
## 66111	-0.1656452	0.0907719	0.1506930	1.812e-01
## 6627	0.6189097	0.4187046	-0.0389922	-5.756e-01
## 6634	0.0485553	0.3388012	0.1611067	-2.550e-01
## 6644	0.0098911	-0.2630645	-0.3615846	-5.683e-01
## 6654	0.4021982	0.2778123	-0.1358907	-1.863e-01
## 6664	0.1104170	0.1722392	-0.4143285	1.840e-01
## 6674	-0.4421972	0.1184143	-0.3812743	-3.722e-01
## 6684	0.4143392	0.3089470	0.0633114	-3.648e-02
## 6694	0.1072259	-0.3087300	-0.8432562	-9.299e-02
## 6704	-0.4535735	0.3902051	-0.1564359	-1.005e-01
## 67111	0.2391657	-0.0039216	-0.0512688	2.898e-01
## 6726	0.3045470	-0.2494710	0.3333226	1.565e-01
## 6733	-0.3803952	-0.0700362	-0.2764642	-1.828e-01
## 6743	-0.1938992	0.2407342	0.1723010	-8.197e-02
## 6753	0.2478062	-0.2296037	-0.5682819	-1.492e-01
## 6763	0.1826436	0.0660646	0.2765318	-5.617e-02
## 6773	0.1293049	-0.2139758	0.1652870	3.560e-01
## 6783	0.1286962	-0.4346062	-0.4210152	2.764e-01
## 6793	0.0710888	-0.1354780	0.2778622	3.736e-01
## 6803	0.1375351	-0.1845472	-0.5133688	1.178e-01

```
# Figure
par(mar=c(4,4,2,2), bty="n")
plot(test.pca_gull1, scaling = scl, choices=c(1,2), type="n", main="", ylim=c(-3,3), xlim=c(-3.5,3.5), xlab="PC1 = 39%", ylab="PC2 = 30%", cex.lab=2, cex.axis=2)
with(data_gull, points(test.pca_gull1, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_gull1, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1]+0.5, y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```

# Test for proportion of variance explained per axe
sum_all_gull <- summary(test.pca_gull1)

prop_gull <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  fec = c(1:4),
  adult = c(1:4),
  breed = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_gull[j,i+1] <- as.numeric(abs(sum_all_gull$species[i,j])/sum(abs(sum_all_gull$species
[i,])))
  }
}

prop_gull

```

```

##      PC      prim      fec      adult      breed
## 1  1 0.21615286 0.03916462 0.50373342 0.49067545
## 2  2 0.35726786 0.51982222 0.16521978 0.05346972
## 3  3 0.34875133 0.30824483 0.06288123 0.21751177
## 4  4 0.07782795 0.13276832 0.26816558 0.23834306

```

```

# Covariances
test <- as.matrix(scale(log(data_gull[,c(2,3,4,7)])))
cov(test)

```

```

##           age_first_repro lifetime_adult breed_freq    fec_one
## age_first_repro      1.0000000    -0.22406312 -0.1550696 -0.09148750
## lifetime_adult      -0.2240631      1.00000000 -0.1348157  0.07337074
## breed_freq          -0.1550696    -0.13481570  1.0000000  0.50642357
## fec_one             -0.0914875      0.07337074  0.5064236  1.00000000

```

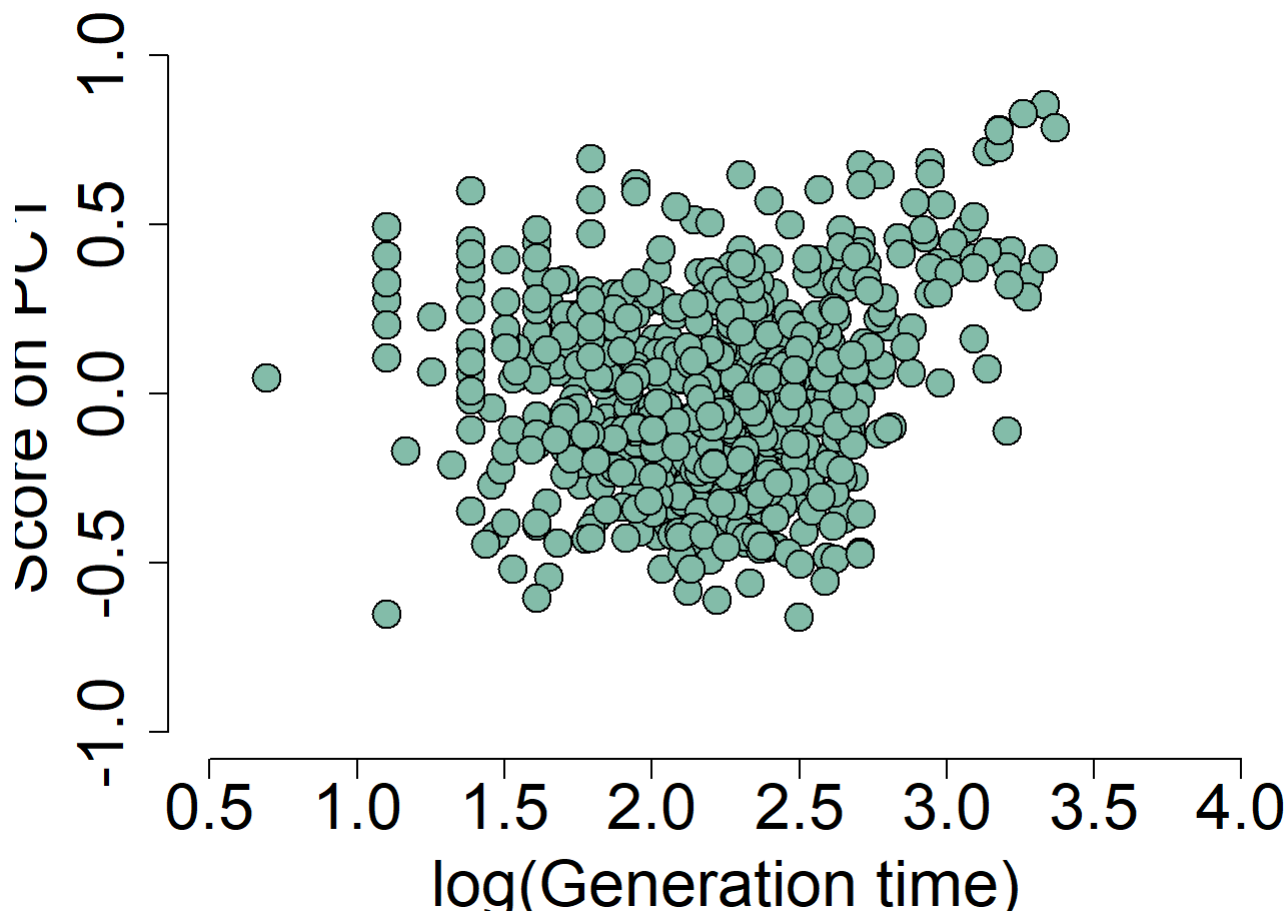
```

# Correlation with generation time
individual_score_gull <- as.numeric(vegan::scores(test.pca_gull1)$sites[,1])
summary(lm(individual_score_gull ~ log(data_gull$generation_time)))

```

```
##
## Call:
## lm(formula = individual_score_gull ~ log(data_gull$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.69054 -0.20808 -0.02598  0.20550  0.73856
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.22857    0.05563  -4.109 4.47e-05 ***
## log(data_gull$generation_time)  0.10381    0.02481   4.184 3.24e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2738 on 677 degrees of freedom
## Multiple R-squared:  0.02521,    Adjusted R-squared:  0.02377
## F-statistic: 17.51 on 1 and 677 DF,  p-value: 3.242e-05
```

```
par(bty="n")
plot(log(data_gull$generation_time), individual_score_gull, pch=21, xlab="log(Generation time)",
     ylab="Score on PC1", xlim=c(0.5,4), ylim=c(-1,1), bg=colors[12], cex=2, cex.axis=2, cex.lab=2)
```



Humans

```
data_humans <- data[which(data$species == "human"),]  
data_humans$fec_one <- data_humans$fecundity + 1  
test.pca_humans1 <- rda(log(data_humans[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_humans1)
```

```
##
## Call:
## rda(X = log(data_humans[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained    4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1      PC2      PC3      PC4
## Eigenvalue      1.724 1.0886 0.9388 0.24863
## Proportion Explained 0.431 0.2721 0.2347 0.06216
## Cumulative Proportion 0.431 0.7032 0.9378 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 5.629028
##
##
## Species scores
##
##           PC1      PC2      PC3      PC4
## age_first_repro 0.4315 -1.73920 2.1690 0.07599
## lifetime_adult -2.5983 0.09785 0.4769 -0.96618
## breed_freq      2.5896 0.49240 0.1881 -0.96834
## fec_one         -0.1169 2.31222 1.5713 0.30426
##
##
## Site scores (weighted sums of species scores)
##
##           PC1      PC2      PC3      PC4
## 1308 -0.238936 0.2416092 1.181e-01 -0.282546
## 2120 -0.148525 0.3800640 2.008e-01 0.012732
## 3109 0.065569 0.7233732 3.188e-01 -0.554169
## 4120 0.021711 0.4350941 3.620e-01 -0.625472
## 5109 -0.106090 0.2993480 3.903e-01 -0.297890
## 6118 -0.141904 0.6299697 1.409e-01 -0.466692
## 7109 -0.183695 0.4249012 -1.858e-02 -0.370303
## 8109 -0.292399 0.4859649 -1.029e-01 -0.183541
## 9108 -0.137955 0.2938565 2.708e-01 -0.349008
## 10109 -0.252949 0.2950080 1.582e-01 -0.124431
## 11108 -0.046286 0.3436802 4.154e-01 -0.339384
## 12104 -0.238228 -0.2553811 -3.040e-01 0.387626
## 1325 0.827773 -0.2201963 -3.384e-01 -0.209190
## 1420 -0.092367 -0.3519076 -1.818e-02 -0.251823
## 1525 -0.120978 -0.4414854 -8.636e-02 -0.193503
## 1620 -0.118565 -0.0206902 -5.720e-01 0.620847
## 1725 -0.263648 -0.0349460 3.764e-01 -0.049594
```

##	1825	-0.261393	-0.1533256	-4.624e-01	-0.495506
##	1920	-0.194326	-0.1946428	5.334e-01	0.438118
##	2020	-0.396930	-0.3117027	8.485e-01	0.586194
##	2124	-0.210799	0.3682221	-7.095e-02	-0.018978
##	2220	0.208217	0.7869418	-3.097e-01	0.579902
##	2320	0.368932	-0.4306910	-6.080e-02	-0.219272
##	2420	-0.318744	-0.1884521	-4.933e-01	-0.090860
##	2520	-0.133551	-0.5555191	-9.327e-02	0.703919
##	2620	-0.234483	0.0895426	-4.285e-01	-0.293458
##	2720	-0.200709	-0.4239538	3.664e-01	-0.007729
##	2820	0.130865	0.0364386	-2.682e-01	0.105821
##	2920	0.755968	1.0838698	1.490e-01	0.356551
##	3020	-0.348639	0.1705774	-2.112e-01	0.242161
##	3120	0.951394	-0.2809322	-2.993e-01	0.098111
##	3220	-0.344751	0.0532110	1.017e-01	0.144847
##	3320	0.915244	-0.0501751	-6.330e-01	0.053966
##	3420	0.809735	-0.1050569	-5.049e-01	-0.231216
##	3520	0.030483	0.5001847	2.636e-01	-0.047845
##	3619	0.221243	-0.0160317	-9.413e-02	0.936523
##	3719	0.034697	0.7321738	1.864e-01	-0.429109
##	3820	-0.115799	0.2595746	8.488e-02	0.377888
##	3919	0.392536	0.3680789	3.422e-01	-0.184335
##	4020	0.116582	0.5013923	1.071e-01	0.116503
##	4123	-0.233349	-0.5819992	2.379e-01	0.230402
##	4220	-0.305998	-0.1542041	-6.100e-02	-0.177511
##	4320	0.074892	0.0372668	-3.710e-01	-0.652291
##	4420	-0.256968	-0.2903451	-3.055e-01	-0.262456
##	4520	-0.014434	-0.4240067	-9.120e-02	-0.260403
##	4619	-0.177644	-0.5572160	2.462e-01	0.242588
##	4719	-0.361825	-0.4233173	-3.441e-02	0.107022
##	4819	-0.450536	-0.1647649	-1.901e-01	0.415630
##	4919	-0.350891	0.3744356	-3.387e-01	-0.140310
##	5019	-0.236820	0.1654234	-3.631e-01	-0.409085
##	5119	-0.209648	0.4410385	-3.288e-01	0.299506
##	5219	-0.167012	-0.1031279	5.995e-01	0.492690
##	5319	1.020456	-0.7217701	3.382e-01	0.182446
##	5419	0.234563	-0.3023449	-2.916e-01	-0.054127
##	5519	-0.327599	-0.2823929	8.004e-02	-0.128796
##	5619	0.101872	0.1436987	-1.762e-01	0.100688
##	5719	-0.310338	0.0260094	-3.143e-01	-0.338660
##	5819	-0.433947	0.2936542	-2.379e-01	0.071410
##	5919	-0.088897	-0.2180189	-4.804e-02	0.273800
##	6019	-0.330146	0.1253979	-3.911e-01	0.003777
##	6119	0.102582	0.0908903	3.874e-01	0.438621
##	6218	0.622320	0.2878289	-1.077e-02	-0.436677
##	6318	0.001990	0.1156516	-2.095e-01	0.976620
##	6418	-0.272041	0.1004968	-1.440e-01	0.315618
##	6517	0.605846	-0.2603223	-1.828e-01	-0.792877
##	6617	0.731738	0.4614626	9.264e-02	-0.035687
##	6717	0.083984	0.5893656	5.337e-02	-0.179997
##	6817	0.291376	0.4917595	3.187e-01	0.376523
##	6917	0.161532	0.5080366	1.077e-01	-0.003416

##	7016	-0.074621	0.2662456	2.184e-01	0.155918
##	7116	-0.069547	0.2770070	3.672e-01	-0.360007
##	7216	0.049853	0.4483063	-8.434e-02	0.716197
##	7316	-0.231589	-0.2349272	6.103e-01	0.332913
##	7416	-0.055976	0.1965778	6.421e-01	-0.396068
##	7516	0.098512	0.5526894	3.059e-01	-0.439202
##	7616	-0.231011	-0.3183976	1.339e-01	0.199498
##	7716	-0.351651	0.2497483	-1.137e-02	0.397993
##	7816	-0.013894	0.3909337	2.971e-02	0.538696
##	7916	-0.248394	-0.4160784	6.291e-02	0.089549
##	8016	0.136584	-0.4716607	6.269e-01	0.220753
##	8116	-0.324671	0.0760669	-2.367e-02	0.462404
##	8215	-0.258727	-0.0005788	2.832e-01	0.131801
##	8316	0.021546	-0.3855129	-5.033e-02	-0.750149
##	8416	-0.348024	0.1361770	-1.009e-01	0.332561
##	8515	-0.335538	-0.0035506	-3.506e-01	0.161579
##	8616	-0.106934	-0.3987684	5.552e-02	-0.106669
##	8715	-0.251337	-0.2053955	2.326e-01	0.084486
##	8816	-0.142814	0.2667454	-4.337e-02	0.514303
##	8916	-0.187800	0.4230261	-1.631e-01	-0.500900
##	9016	-0.258142	0.2953178	1.599e-01	-0.137822
##	9116	-0.239972	0.2369917	3.593e-01	-0.080179
##	9215	-0.230394	0.2379276	2.349e-01	-0.078237
##	9315	0.461326	-0.1856678	-5.196e-01	0.323733
##	9414	0.003854	0.0884962	4.406e-01	-0.573786
##	9514	0.007259	0.2473510	2.070e-01	-0.416564
##	9614	0.479364	-0.3008072	-3.531e-01	0.345760
##	9715	-0.297777	0.3531938	-2.048e-01	-0.285212
##	9815	-0.098641	-0.3775939	9.238e-01	0.094897
##	9915	-0.187096	0.1261142	-1.387e-01	0.267919
##	10015	0.355451	0.4709236	1.652e-01	0.247032
##	10115	0.261516	0.0648495	-8.875e-01	0.243042
##	10215	0.942956	-0.2270662	-3.772e-01	0.087806
##	10315	-0.444357	-0.2261382	1.487e-02	0.310892
##	10415	0.818959	-0.1639355	-4.197e-01	-0.219953
##	10515	0.271644	-0.3045565	-3.041e-01	0.041485
##	10615	0.002432	-0.0601524	6.492e-02	0.144054
##	10715	-0.337885	-0.0358677	3.269e-02	0.037693
##	10815	-0.220701	0.2013504	-6.924e-02	0.198467
##	10914	-0.208860	-0.0437413	1.316e-01	-0.183941
##	11014	-0.366001	0.1693217	-2.164e-01	0.463066
##	11114	-0.505705	-0.0376496	-4.964e-01	0.273253
##	11214	-0.206288	-0.2669129	-1.723e-01	-0.145479
##	11313	0.031298	-0.0386782	-7.609e-01	0.287544
##	11414	-0.312199	0.1094221	2.671e-01	0.041611
##	11514	-0.222824	-0.1709281	5.679e-01	0.158550
##	11614	-0.328682	0.1863488	7.532e-02	0.183581
##	11714	-0.437373	0.0330619	-3.079e-01	0.360749
##	11814	-0.066856	0.3777380	3.256e-01	-0.394517
##	11914	-0.101619	0.2921542	2.312e-04	-0.382454
##	12014	-0.148970	-0.5363122	6.908e-02	-0.286393
##	12114	-0.214403	-0.2743662	1.765e-01	-0.103220

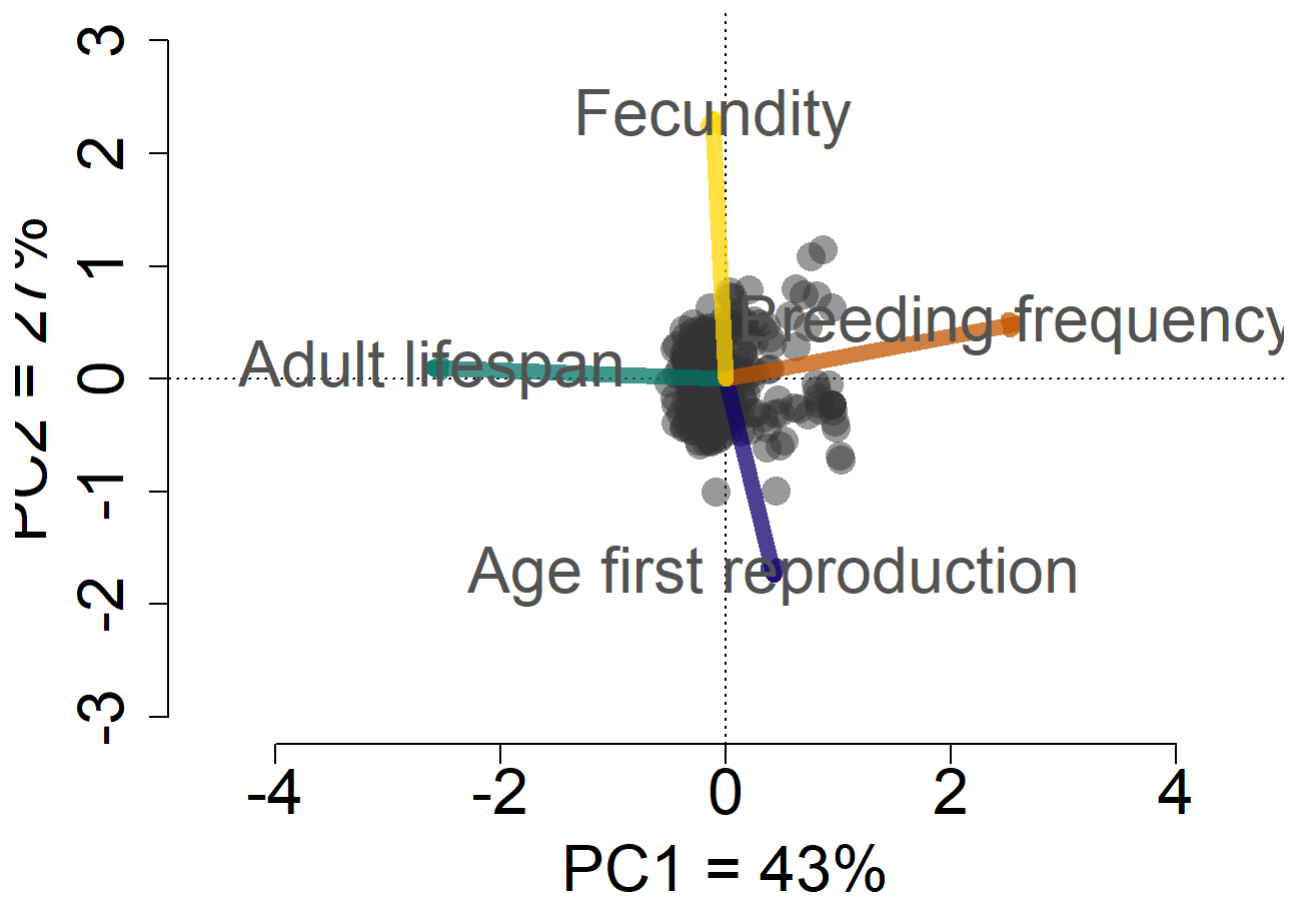
##	12214	0.182611	0.0126023	-7.333e-01	-0.186698
##	12314	0.942956	-0.2270662	-3.772e-01	0.087806
##	12413	-0.050254	-0.2954853	3.479e-01	-0.290396
##	12514	-0.099695	-0.4161414	6.463e-01	-0.261157
##	12614	0.160153	0.0640930	3.563e-03	0.328813
##	12714	-0.250467	-0.1465313	-1.928e-01	-0.226304
##	12814	-0.133686	-0.5523837	4.416e-02	-0.029686
##	12912	-0.070012	-0.5322487	-7.940e-02	0.479881
##	1309	-0.397026	0.1038214	2.923e-02	-0.001399
##	13111	0.043530	-0.1630732	-1.073e-02	-0.285531
##	1326	-0.236178	-0.5462707	7.871e-02	-0.293958
##	1335	-0.049054	-0.5128185	7.381e-02	-0.450934
##	1345	0.014918	0.0706259	-8.161e-01	-0.264782
##	1355	-0.089359	-1.0009334	7.466e-01	-0.231652
##	1365	-0.130308	-0.5121668	2.963e-01	-0.423704
##	1375	0.028895	-0.4629407	-1.467e-02	-0.343089
##	1385	-0.182379	-0.1945913	2.971e-05	-0.031209
##	1395	-0.227530	0.0393812	-1.764e-01	-0.412066
##	1405	-0.086051	0.0787775	8.656e-02	-0.306017
##	14111	-0.168900	0.0549455	-3.936e-02	-0.457811
##	1425	-0.033946	-0.3969828	3.495e-01	0.146519
##	1435	1.014537	-0.6839864	2.836e-01	0.175218
##	1445	-0.341822	-0.3875343	5.060e-01	0.444311
##	1455	-0.276545	-0.1186465	-1.916e-01	0.524576
##	1465	-0.441841	-0.3932248	7.006e-01	0.656344
##	1475	-0.308485	-0.4506945	3.672e-01	-0.029656
##	1485	-0.085869	-0.1734989	5.471e-01	-0.083940
##	1495	0.149052	0.2232760	-1.209e-01	0.076152
##	1505	0.208774	0.3615355	-1.688e-01	0.238350
##	15111	-0.297282	0.2143526	-8.236e-01	0.007865
##	1526	0.942956	-0.2270662	-3.772e-01	0.087806
##	1535	0.188419	0.1181422	-9.030e-02	0.413661
##	1545	-0.029944	0.5189172	1.779e-02	-0.129689
##	1555	-0.329880	0.0751032	-1.537e-01	0.339767
##	1565	0.098943	-0.2359922	-4.913e-01	0.420480
##	1575	-0.292263	-0.0721580	1.162e-01	0.079627
##	1585	0.216767	-0.4529932	2.758e-01	-0.454213
##	1594	-0.139697	-0.2602090	6.024e-02	-0.122913
##	1605	-0.455318	0.2662630	-7.478e-01	-0.072383
##	16111	-0.139342	-0.2583500	-2.943e-01	-0.518328
##	1625	0.152134	-0.1834401	5.545e-02	0.113741
##	1634	0.974748	-0.4300055	-8.370e-02	0.126630
##	1645	0.068376	-0.3295914	1.003e-01	-0.796077
##	1655	0.011437	0.0368138	1.517e-01	-0.163426
##	1665	0.729123	-0.3188591	-1.469e-01	-0.485996
##	1675	-0.284612	0.1044205	-4.087e-01	-0.231944
##	1685	-0.370263	0.4347349	-1.386e-01	0.102835
##	1695	0.252485	0.3309285	7.703e-02	0.143801
##	1705	-0.197104	0.1767076	2.999e-01	-0.007727
##	17111	0.005022	-0.0330531	1.951e-01	0.405799
##	1726	0.151021	0.2466263	4.146e-01	-0.289021
##	1734	0.033284	0.7754989	-2.918e-01	0.797986

## 1745	0.338433	-0.3447330	5.920e-02	-1.503937
## 1755	0.448289	-0.9928069	7.072e-01	0.116705
## 1765	-0.225020	-0.3442534	-2.271e-01	-0.245541
## 1775	0.124370	-0.3298140	-2.989e-01	0.149685
## 1785	-0.076138	-0.0589399	2.146e-01	-0.283415
## 1795	-0.186188	-0.1612586	-3.502e-01	-0.276639
## 1804	-0.199779	0.0231595	-3.429e-01	-0.217780
## 18111	-0.351943	0.1799664	-7.078e-02	0.498241
## 1826	0.407062	0.3033799	1.479e+00	-0.715090
## 1835	0.439565	-0.3108408	6.087e-01	0.160578
## 1845	-0.196200	0.2777798	6.014e-03	-0.339431
## 1855	0.164112	0.4862850	8.390e-02	0.163404
## 1865	0.940826	0.6322496	7.534e-01	0.738631
## 1875	0.836212	-0.2740623	-2.605e-01	-0.198885
## 1884	-0.321646	0.0266839	-3.105e-01	-0.367819
## 1894	0.388837	0.0912655	1.008e-01	0.391192
## 1904	0.178557	-0.1339743	-5.512e-01	-0.014604
## 19111	-0.192730	-0.0499100	-4.072e-01	-0.094478
## 1924	-0.153121	-0.0133710	-2.362e-01	-0.334483
## 1934	-0.339115	0.1229813	-5.345e-01	-0.372766
## 1944	-0.191387	-0.1721002	2.340e-01	0.466104
## 1954	-0.281065	0.1310468	-3.826e-01	-0.388021
## 1964	-0.135912	-0.1756321	7.697e-03	-0.079987
## 1974	-0.162754	-0.2397706	4.085e-01	-0.106543
## 1984	-0.282978	0.4167321	-1.650e-01	-0.051499
## 1994	-0.117232	-0.0846314	-6.301e-02	0.400225
## 2004	0.371623	0.0846289	-8.903e-01	0.149305
## 20111	-0.280462	-0.3710358	1.050e-01	0.101676
## 2024	0.653651	-0.2631736	-1.989e-01	-0.669612
## 2034	-0.186976	-0.2820494	2.437e-01	0.765540
## 2044	0.860431	1.1454305	1.121e-02	0.640456
## 2054	-0.080709	0.1359985	-4.370e-01	0.139744
## 2064	-0.147650	0.3763045	1.970e-01	0.030439
## 2074	-0.076287	0.0078868	-2.584e-01	0.349952
## 2084	-0.057985	-0.0808468	-3.249e-01	0.112025
## 2094	0.056377	0.0443532	-1.172e-01	0.242111
## 21010	0.942956	-0.2270662	-3.772e-01	0.087806
## 21111	-0.188362	0.3322872	1.736e-01	-0.077810
## 2125	-0.013108	-0.0744589	-3.215e-01	-0.092513
## 2134	-0.288990	0.2706272	-6.003e-01	-0.120077
## 2144	-0.332092	0.1755409	-3.475e-01	-0.082826
## 2154	-0.374580	-0.4326180	7.765e-02	0.006978
## 2164	-0.281811	-0.4614053	1.458e-01	-0.006775
## 2174	-0.218491	-0.4387275	7.294e-01	0.065326
## 2184	-0.311848	-0.1285792	-3.146e-01	0.168337
## 2194	-0.014365	-0.3033210	-1.376e-01	0.454065
## 2204	-0.322502	0.1823372	-3.421e-01	-0.136426
## 22111	-0.180732	0.2052312	7.979e-02	-0.351956
## 2224	-0.375738	0.2230535	-2.275e-02	0.248530
## 2234	-0.148520	0.4508838	-4.677e-01	-0.575681
## 2244	-0.015507	0.1124806	3.383e-01	-0.235507
## 2254	0.695201	0.7433960	6.901e-01	0.126004

```
## 2264 -0.149918 0.1812152 1.866e-01 -0.454640
## 2274 -0.022675 0.4633958 1.265e-01 -0.515088
## 2284 0.142723 -0.1510820 -5.643e-01 0.168851
## 2294 -0.059398 0.4205073 9.181e-02 -0.298438
## 2304 -0.302007 0.2541697 -1.500e-04 -0.103046
## 23111 -0.356044 -0.2072613 2.393e-01 0.446150
## 2324 -0.076232 -0.1558036 -2.229e-01 -0.568562
## 2334 0.045901 -0.1437128 1.348e-01 -0.638798
## 2344 0.518367 -0.5497755 6.931e-03 0.393389
## 2354 -0.099009 -0.4799523 8.091e-02 0.382226
## 2364 0.800061 -0.0433052 -5.942e-01 -0.243030
## 2374 0.367413 -0.6111173 9.004e-02 0.372359
## 2384 0.810383 0.7365260 6.513e-01 0.423000
## 2394 0.250115 0.5194964 7.704e-02 0.631066
## 2404 0.619729 0.7990814 6.381e-01 -0.057611
## 24111 -0.090972 -0.4215441 -2.123e-01 0.317022
## 2424 0.477520 -0.5913379 8.728e-02 0.278621
## 2434 0.085884 -0.3086370 -2.613e-01 -0.220193
## 2444 -0.233948 -0.0941452 8.713e-02 0.074476
## 2454 -0.367779 0.0236036 8.244e-02 0.148662
## 2464 0.039232 -0.1990396 -4.276e-01 -0.212103
## 2474 -0.174822 0.3188347 3.286e-02 -0.287498
## 2484 -0.144153 -0.1519693 1.256e-01 0.514786
## 2494 -0.072122 0.1356082 2.454e-01 -0.102689
## 2504 0.051866 -0.0227787 -2.620e-02 -0.194902
## 25111 0.967265 -0.3822391 -1.528e-01 0.117492
## 2524 0.572366 0.5646284 1.013e+00 -0.230666
```

Figure

```
par(mar=c(4,4,2,2), bty="n")
plot(test.pca_humans1, scaling = scl, choices=c(1,2), type="n", main="", ylim=c(-3,3), xlim=c(-3.5,3.5), xlab="PC1 = 43%", ylab="PC2 = 27%", cex.axis=2, cex.lab=2)
with(data_humans, points(test.pca_humans1, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_humans1, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```
# Test for proportion of variance explained per axe
sum_all_humans <- summary(test.pca_humans1)

prop_humans <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_humans[j,i+1] <- as.numeric(abs(sum_all_humans$species[i,j])/sum(abs(sum_all_humans$species[i,])))
  }
}

prop_humans
```

```
##      PC      prim      adult      breed      freq
## 1  1 0.09772028 0.62771656 0.61098595 0.02716612
## 2  2 0.39386436 0.02364015 0.11617626 0.53714018
## 3  3 0.49120584 0.11522275 0.04436976 0.36501258
## 4  4 0.01720952 0.23342054 0.22846803 0.07068112
```

Covariances

```
test <- as.matrix(scale(log(data_humans[,c(2,3,4,7)])))
cov(test)
```

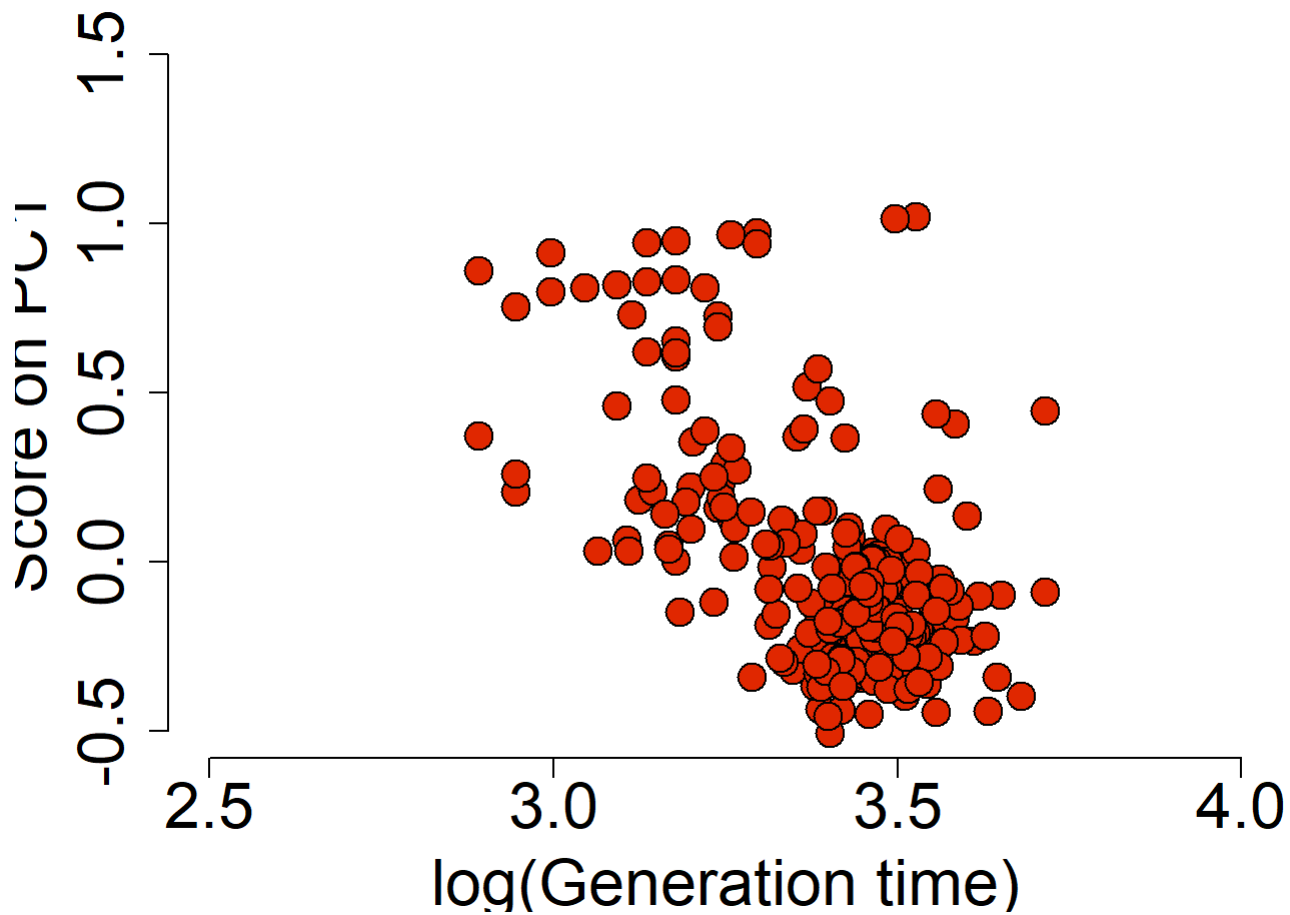
```
##           age_first_repro lifetime_adult breed_freq fec_one
## age_first_repro      1.00000000 -0.04169533 0.07515743 -0.0808719
## lifetime_adult      -0.04169533  1.00000000 -0.71387769 0.1244106
## breed_freq          0.07515743 -0.71387769 1.00000000 0.1056071
## fec_one             -0.08087190 0.12441062 0.10560711 1.0000000
```

Correlation with generation time

```
individual_score_humans <- as.numeric(vegan::scores(test.pca_humans1)$sites[,1])
summary(lm(individual_score_humans~ log(data_humans$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_humans ~ log(data_humans$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.49226 -0.19120 -0.03396  0.11925  1.20814
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)         4.7212     0.3978   11.87  <2e-16 ***
## log(data_humans$generation_time) -1.3921     0.1172  -11.88  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2846 on 250 degrees of freedom
## Multiple R-squared:  0.3609, Adjusted R-squared:  0.3583
## F-statistic: 141.2 on 1 and 250 DF,  p-value: < 2.2e-16
```

```
par(bty="n")
plot(log(data_humans$generation_time), individual_score_humans, pch=21, xlab="log(Generation time)", ylab="Score on PC1", xlim=c(2.5,4), ylim=c(-0.5,1.5), bg=colors[9], cex=2, cex.axis=2, cex.lab=2)
```



Wandering albatross

```
data_WA <- data[which(data$species == "wanderingalbatross"),]  
data_WA$fec_one <- data_WA$fecundity + 1  
test.pca_WA1 <- rda(log(data_WA[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_WA1)
```

```
##
## Call:
## rda(X = log(data_WA[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained    4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1      PC2      PC3      PC4
## Eigenvalue      2.068 0.9921 0.7849 0.15518
## Proportion Explained 0.517 0.2480 0.1962 0.03879
## Cumulative Proportion 0.517 0.7650 0.9612 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 7.21909
##
##
## Species scores
##
##           PC1      PC2      PC3      PC4
## age_first_repro 0.5295 3.54798 0.4003 -0.008546
## lifetime_adult  3.3586 -0.13730 -0.8334 -1.017343
## breed_freq      -3.3074 0.07181 1.0501 -0.990997
## fec_one         -2.1075 0.55988 -2.8756 -0.068221
##
##
## Site scores (weighted sums of species scores)
##
##           PC1      PC2      PC3      PC4
## 1300 -2.870e-02 -1.781e-01 2.095e-01 -0.444451
## 2105 -1.091e-01 -1.228e-01 -1.604e-01 0.229194
## 3104 1.864e-01 2.931e-01 4.368e-02 -0.136113
## 4104 9.422e-02 -1.562e-01 -1.249e-01 0.608720
## 5104 -4.579e-01 4.106e-01 -1.418e-01 0.010175
## 6104 3.639e-01 1.356e-01 -7.633e-02 -0.026503
## 7104 2.257e-01 1.648e-01 -1.483e-01 0.168219
## 8104 -4.680e-01 2.695e-01 -1.619e-01 0.012347
## 9104 2.655e-02 3.469e-02 -1.503e-01 0.173244
## 10104 3.268e-01 -1.843e-01 -1.876e-01 0.412061
## 11104 -2.098e-01 -4.222e-01 8.268e-01 0.148147
## 12100 1.468e-01 1.595e-01 2.359e-03 0.116088
## 1316 -4.680e-01 2.695e-01 -1.619e-01 0.012347
## 1416 2.950e-01 -3.847e-01 -1.738e-01 0.115530
## 1516 1.866e-01 1.726e-01 -1.700e-01 -0.032839
## 1616 -9.659e-02 5.153e-02 -1.355e-01 0.226510
## 1716 1.918e-01 1.346e-01 1.641e-01 0.135495
```

## 1816	4.117e-02	1.756e-01	-4.897e-03	-0.085864
## 1915	1.232e-01	-1.618e-01	-1.294e-01	-0.141564
## 2015	1.644e-02	1.425e-01	3.108e-01	-0.564740
## 2115	1.249e-02	-2.020e-01	3.705e-01	-0.267907
## 2215	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 2315	-4.579e-01	4.106e-01	-1.418e-01	0.010175
## 2415	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 2515	1.373e-01	-3.976e-02	3.586e-01	-0.037173
## 2614	1.156e-01	-1.866e-01	9.760e-02	-0.251759
## 2714	-3.387e-01	2.585e-01	-2.464e-01	-0.509707
## 2814	1.254e-01	4.518e-01	-1.055e-01	0.094378
## 2914	2.301e-01	1.624e-01	-1.419e-01	-0.258535
## 3014	3.429e-01	-2.291e-01	1.937e-01	0.659098
## 3114	1.800e-01	5.648e-01	-1.025e-01	0.023651
## 3214	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 3314	2.371e-01	-3.565e-03	-9.876e-02	-0.248295
## 3414	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 3514	2.552e-01	1.485e-01	-4.239e-02	0.180927
## 3614	2.021e-01	-1.654e-01	-1.954e-01	0.173304
## 3714	1.494e-01	-3.333e-02	2.913e-01	0.226233
## 3814	2.289e-01	-1.288e-02	-8.444e-05	-0.007384
## 3914	2.122e-01	-9.223e-03	-2.305e-02	-0.599153
## 4014	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 4114	1.232e-01	-1.707e-01	-4.473e-02	0.121173
## 4214	-8.031e-02	2.397e-01	8.688e-01	-0.191532
## 4314	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 4414	2.047e-01	-1.382e-02	4.958e-02	0.435056
## 4514	3.915e-02	7.931e-03	6.086e-02	-0.266878
## 4614	2.173e-01	5.523e-03	-1.579e-01	-0.473628
## 4714	1.544e-01	1.403e-02	-1.312e-01	0.413361
## 4814	-1.745e-02	-3.890e-01	2.735e-01	-0.054889
## 4914	-1.232e-01	-3.204e-01	-1.885e-01	0.232237
## 5014	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 5114	1.529e-02	-2.426e-01	7.171e-01	-0.706330
## 5214	2.472e-01	-3.301e-02	1.511e-01	-0.148653
## 5314	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 5414	3.772e-02	1.906e-01	-1.280e-01	0.170843
## 5514	2.586e-01	-1.874e-01	-8.437e-02	-0.273790
## 5614	3.505e-01	1.459e-01	-1.405e-01	0.406976
## 5714	1.747e-01	-3.695e-01	-1.522e-01	-0.013544
## 5814	2.802e-01	-1.190e-03	-1.651e-01	0.295596
## 5914	7.109e-02	-1.712e-01	2.983e-02	0.200524
## 6014	6.834e-02	-3.458e-01	-2.192e-01	0.106660
## 6113	2.122e-01	-5.742e-03	-5.241e-02	-0.527878
## 6213	-8.116e-02	-1.483e-01	2.184e-02	-0.091514
## 6313	1.876e-01	1.511e-01	1.623e-02	-0.179232
## 6413	2.316e-01	-1.817e-01	-8.948e-02	0.186012
## 6512	2.913e-01	-1.075e-02	-1.062e-01	-0.221931
## 6612	3.372e-01	-3.276e-02	3.985e-02	0.320192
## 6712	-9.659e-02	5.153e-02	-1.355e-01	0.226510
## 6812	2.721e-01	2.793e-01	4.816e-02	-0.347768
## 6912	3.450e-01	-1.858e-01	-1.995e-01	0.338810

##	7012	2.830e-01	5.450e-01	-6.629e-02	-0.137753
##	7112	1.807e-01	-2.135e-02	1.418e-01	0.137895
##	7212	2.105e-01	-1.201e-02	7.591e-03	-0.422063
##	7312	1.917e-01	5.414e-01	8.704e-02	-0.089853
##	7412	3.189e-01	1.395e-01	-4.381e-02	0.305083
##	7512	-4.792e-01	1.136e-01	-1.841e-01	0.014748
##	7612	-2.870e-02	-1.781e-01	2.095e-01	-0.444451
##	7712	2.159e-01	1.421e-01	7.181e-02	0.432655
##	7812	1.411e-01	2.063e-02	-1.801e-01	0.032031
##	7912	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
##	8012	1.602e-01	-1.712e-01	-9.122e-02	0.048365
##	8112	1.407e-02	-1.396e-01	-1.751e-01	0.175928
##	8312	-1.832e-01	-5.030e-02	8.798e-01	0.142420
##	8412	1.995e-01	1.671e-01	-1.311e-01	0.274255
##	8612	-4.487e-01	5.393e-01	-1.234e-01	0.008192
##	8711	-4.086e-02	8.748e-03	1.588e-01	-0.298984
##	8812	1.372e-01	4.438e-01	-3.930e-02	0.599480
##	8912	-4.579e-01	4.106e-01	-1.418e-01	0.010175
##	9012	1.411e-01	2.063e-02	-1.801e-01	0.032031
##	9112	-9.659e-02	5.153e-02	-1.355e-01	0.226510
##	9211	2.160e-01	-5.855e-04	-9.467e-02	-0.129060
##	9311	2.692e-01	-1.265e-03	-1.471e-01	0.431048
##	9411	1.232e-01	-1.618e-01	-1.294e-01	-0.141564
##	9511	2.257e-01	-1.692e-01	-2.025e-01	-0.287966
##	9611	2.360e-01	7.079e-03	-1.846e-01	0.083981
##	9712	1.721e-01	-1.993e-01	1.550e-01	0.642302
##	9812	3.766e-02	5.429e-01	2.846e-01	0.259232
##	9912	1.522e-01	1.765e-01	-1.579e-01	0.029630
##	10012	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
##	10112	3.467e-01	5.277e-01	1.155e-02	0.299104
##	10212	-8.543e-02	2.074e-01	-1.133e-01	0.224109
##	10312	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
##	10412	-4.487e-01	5.393e-01	-1.234e-01	0.008192
##	10512	7.109e-02	-1.712e-01	2.983e-02	0.200524
##	10612	1.165e-01	-1.794e-01	3.608e-02	-0.090000
##	10712	8.248e-02	-1.482e-01	-1.910e-01	0.103618
##	10812	-1.277e-01	-1.857e-01	4.058e-01	-0.554756
##	10912	-4.792e-01	1.136e-01	-1.841e-01	0.014748
##	11012	1.427e-01	-3.618e-01	-1.818e-01	-0.176508
##	11112	1.252e-02	-3.640e-01	7.827e-03	-0.261151
##	11212	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
##	11311	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
##	11412	2.315e-01	-1.735e-01	-1.687e-01	-0.126689
##	11512	3.377e-03	-3.522e-01	-8.016e-02	-0.077736
##	11612	-7.432e-05	-3.372e-01	-2.033e-01	0.178971
##	11712	-2.747e-01	3.862e-02	2.041e-01	-0.261643
##	11812	8.248e-02	-1.482e-01	-1.910e-01	0.103618
##	11912	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
##	12012	-4.792e-01	1.136e-01	-1.841e-01	0.014748
##	12112	-1.378e-02	-1.755e-01	1.822e-01	0.270295
##	12212	-3.013e-01	-3.333e-01	1.511e-01	-0.255916
##	12312	-8.116e-02	-1.483e-01	2.184e-02	-0.091514

```
## 12411 -4.916e-01 -6.070e-02 -2.090e-01 0.017432
## 12512 1.602e-01 -1.712e-01 -9.122e-02 0.048365
## 12612 -4.916e-01 -6.070e-02 -2.090e-01 0.017432
## 12712 -1.378e-02 -1.755e-01 1.822e-01 0.270295
## 12812 -2.747e-01 3.862e-02 2.041e-01 -0.261643
## 12910 -4.916e-01 -6.070e-02 -2.090e-01 0.017432
## 1303 7.109e-02 -1.712e-01 2.983e-02 0.200524
## 1317 2.542e-01 5.532e-03 -1.964e-01 0.010730
## 1323 -1.378e-02 -1.755e-01 1.822e-01 0.270295
## 1333 4.117e-02 1.756e-01 -4.897e-03 -0.085864
## 1343 -4.916e-01 -6.070e-02 -2.090e-01 0.017432
## 1353 -2.098e-01 -4.222e-01 8.268e-01 0.148147
## 1363 2.021e-01 -1.654e-01 -1.954e-01 0.173304
## 1373 -2.098e-01 -4.222e-01 8.268e-01 0.148147
## 1383 3.162e-03 4.303e-02 -2.007e-01 -0.176121
## 1393 2.021e-01 -1.654e-01 -1.954e-01 0.173304
## 1403 1.369e-01 -2.076e-01 2.664e-01 0.228917
## 1417 -6.272e-02 4.864e-02 -1.576e-01 0.089805
## 1423 -1.232e-01 -3.204e-01 -1.885e-01 0.232237
## 1433 7.109e-02 -1.712e-01 2.983e-02 0.200524
## 1443 8.327e-02 -1.455e-01 -2.204e-01 -0.103385
## 1453 -5.058e-01 -2.583e-01 -2.371e-01 0.020475
## 1463 7.109e-02 -1.712e-01 2.983e-02 0.200524
## 1473 -5.221e-01 -4.864e-01 -2.697e-01 0.023988
## 1483 2.472e-01 1.630e-01 -1.623e-01 0.081580
## 1493 1.407e-02 -1.396e-01 -1.751e-01 0.175928
## 1503 -1.141e-01 -2.315e-01 8.016e-01 -0.184275
## 1517 -1.091e-01 -1.228e-01 -1.604e-01 0.229194
## 1523 2.451e-01 -6.059e-01 -2.269e-01 -0.035645
## 1533 -5.058e-01 -2.583e-01 -2.371e-01 0.020475
## 1543 -1.091e-01 -1.228e-01 -1.604e-01 0.229194
## 1553 -6.868e-02 2.604e-02 4.669e-02 -0.094198
## 1563 -4.792e-01 1.136e-01 -1.841e-01 0.014748
## 1573 -1.957e-01 -2.246e-01 8.550e-01 0.145104
## 1583 -2.098e-01 -4.222e-01 8.268e-01 0.148147
## 1592 -5.221e-01 -4.864e-01 -2.697e-01 0.023988
## 1603 -4.916e-01 -6.070e-02 -2.090e-01 0.017432
## 1617 -6.868e-02 2.604e-02 4.669e-02 -0.094198
## 1623 6.269e-02 -5.857e-01 -1.672e-01 -0.155441
## 1632 -5.221e-01 -4.864e-01 -2.697e-01 0.023988
## 1643 1.030e-01 1.714e-01 -5.039e-02 -0.111112
## 1653 -1.232e-01 -3.204e-01 -1.885e-01 0.232237
## 1663 -4.916e-01 -6.070e-02 -2.090e-01 0.017432
## 1673 -1.663e-03 -3.996e-01 3.423e-01 -0.264864
## 1683 -2.033e-01 -3.652e-01 3.007e-01 -0.426896
## 1693 1.781e-01 -3.857e-01 -3.449e-03 0.440783
## 1703 -4.680e-01 2.695e-01 -1.619e-01 0.012347
## 1717 8.248e-02 -1.482e-01 -1.910e-01 0.103618
## 1723 -7.432e-05 -3.372e-01 -2.033e-01 0.178971
## 1732 -2.942e-02 -3.515e-01 -4.941e-02 -0.354398
## 1743 -3.013e-01 -3.333e-01 1.511e-01 -0.255916
## 1753 3.281e-01 -1.861e-01 -1.702e-01 0.571378
```

## 1763	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 1773	1.357e-01	3.563e-03	-1.987e-02	0.118489
## 1783	1.584e-01	-2.095e-01	2.524e-01	0.142279
## 1793	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 1802	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 1817	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 1823	5.367e-02	-3.801e-01	1.006e-01	-0.056041
## 1833	-6.293e-02	-6.642e-01	6.876e-01	-0.507099
## 1843	-2.793e-02	-3.731e-01	1.540e-01	0.273338
## 1853	2.234e-01	-1.018e-02	-3.039e-02	-0.644444
## 1863	7.931e-02	-1.588e-01	-9.747e-02	-0.106027
## 1873	-1.957e-01	-2.246e-01	8.550e-01	0.145104
## 1882	-1.641e-02	-5.653e-01	-2.358e-01	0.182484
## 1892	2.484e-01	-3.779e-01	-1.746e-01	0.025444
## 1902	-3.177e-01	-5.614e-01	1.186e-01	-0.252403
## 1916	1.407e-02	-1.396e-01	-1.751e-01	0.175928
## 1922	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
## 1932	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 1942	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 1952	-1.232e-01	-3.204e-01	-1.885e-01	0.232237
## 1962	9.179e-02	1.551e-02	-7.262e-02	-0.108711
## 1972	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 1982	1.407e-02	-1.396e-01	-1.751e-01	0.175928
## 1992	2.120e-01	-1.727e-01	-1.489e-01	-0.074570
## 2002	1.461e-01	-3.688e-01	-1.194e-01	0.051408
## 2016	8.248e-02	-1.482e-01	-1.910e-01	0.103618
## 2022	1.232e-01	-1.707e-01	-4.473e-02	0.121173
## 2032	1.772e-01	-3.576e-01	-2.638e-01	-0.139310
## 2042	1.979e-01	-3.703e-01	-1.771e-01	-0.071527
## 2052	1.758e-01	-1.632e-01	-1.782e-01	0.279340
## 2062	8.248e-02	-1.482e-01	-1.910e-01	0.103618
## 2072	2.338e-01	-3.783e-01	-1.498e-01	0.100236
## 2082	8.602e-02	-1.751e-01	3.225e-02	-0.395073
## 2092	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 2106	6.834e-02	-3.458e-01	-2.192e-01	0.106660
## 2116	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 2122	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 2132	-8.116e-02	-1.483e-01	2.184e-02	-0.091514
## 2142	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
## 2152	-1.957e-01	-2.246e-01	8.550e-01	0.145104
## 2162	-1.378e-02	-1.755e-01	1.822e-01	0.270295
## 2172	1.407e-02	-1.396e-01	-1.751e-01	0.175928
## 2182	-1.232e-01	-3.204e-01	-1.885e-01	0.232237
## 2192	3.661e-01	-4.115e-01	-2.369e-02	0.438158
## 2202	1.232e-01	-1.618e-01	-1.294e-01	-0.141564
## 2216	2.146e-01	8.908e-03	-1.705e-01	0.170620
## 2222	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
## 2232	-1.378e-02	-1.755e-01	1.822e-01	0.270295
## 2242	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 2252	-6.868e-02	2.604e-02	4.669e-02	-0.094198
## 2262	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 2272	-4.579e-01	4.106e-01	-1.418e-01	0.010175

##	2282	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
##	2292	2.013e-01	2.379e-03	-9.915e-02	-0.019271
##	2302	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
##	2316	-2.635e-01	1.945e-01	2.263e-01	-0.264044
##	2322	6.834e-02	-3.458e-01	-2.192e-01	0.106660
##	2332	1.248e-01	-2.141e-01	3.338e-01	-0.034489
##	2342	-7.432e-05	-3.372e-01	-2.033e-01	0.178971
##	2352	-1.378e-02	-1.755e-01	1.822e-01	0.270295
##	2362	6.723e-02	-3.652e-01	-5.344e-02	-0.263459
##	2372	8.248e-02	-1.482e-01	-1.910e-01	0.103618
##	2382	-8.116e-02	-1.483e-01	2.184e-02	-0.091514
##	2392	-1.141e-01	-2.315e-01	8.016e-01	-0.184275
##	2402	-2.196e-01	-5.933e-01	2.681e-01	-0.423383
##	2416	1.035e-01	-3.902e-01	1.178e-01	-0.378380
##	2422	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
##	2432	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
##	2442	-1.745e-02	-3.890e-01	2.735e-01	-0.054889
##	2452	2.493e-01	-2.197e-01	2.297e-01	0.462335
##	2462	-4.182e-01	9.650e-01	-6.270e-02	0.001637
##	2472	9.422e-02	-1.562e-01	-1.249e-01	0.608720
##	2482	-4.792e-01	1.136e-01	-1.841e-01	0.014748
##	2492	-2.261e-01	-6.503e-01	7.943e-01	0.151660
##	2502	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
##	2516	-1.245e-01	-3.316e-01	-9.923e-02	-0.317853
##	2522	-9.659e-02	5.153e-02	-1.355e-01	0.226510
##	2531	4.782e-02	3.316e-01	-1.079e-01	0.168671
##	2541	1.407e-02	-1.396e-01	-1.751e-01	0.175928
##	2551	1.407e-02	-1.396e-01	-1.751e-01	0.175928
##	2561	-5.751e-02	1.820e-01	6.892e-02	-0.096599
##	2571	-8.116e-02	-1.483e-01	2.184e-02	-0.091514
##	2581	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
##	2591	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
##	2601	-4.579e-01	4.106e-01	-1.418e-01	0.010175
##	2615	-1.832e-01	-5.030e-02	8.798e-01	0.142420
##	2621	7.109e-02	-1.712e-01	2.983e-02	0.200524
##	2631	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
##	2641	-4.579e-01	4.106e-01	-1.418e-01	0.010175
##	2651	-4.402e-01	6.578e-01	-1.065e-01	0.006368
##	2661	-1.141e-01	-2.315e-01	8.016e-01	-0.184275
##	2671	7.931e-02	-1.588e-01	-9.747e-02	-0.106027
##	2681	-1.832e-01	-5.030e-02	8.798e-01	0.142420
##	2691	-4.402e-01	6.578e-01	-1.065e-01	0.006368
##	2701	9.422e-02	-1.562e-01	-1.249e-01	0.608720
##	2715	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
##	2721	6.516e-02	-3.564e-01	-1.256e-01	-0.102984
##	2731	-3.301e-03	-1.914e-01	3.017e-01	-0.057932
##	2741	-8.543e-02	2.074e-01	-1.133e-01	0.224109
##	2751	2.867e-01	2.660e-01	1.453e-01	-0.430289
##	2761	3.945e-01	2.707e-01	-3.562e-02	-0.268075
##	2771	3.562e-01	2.811e-01	-7.861e-02	-0.324395
##	2781	3.123e-01	2.876e-01	-7.603e-02	-0.217583
##	2791	1.649e-01	1.519e-01	2.868e-02	-0.668303

##	2801	2.630e-01	-3.518e-03	-1.389e-01	-0.506426
##	2815	-1.232e-01	-3.204e-01	-1.885e-01	0.232237
##	2821	2.449e-01	1.569e-01	-1.188e-01	-0.564603
##	2831	2.608e-01	2.111e-05	-1.603e-01	-0.176597
##	2851	-5.221e-01	-4.864e-01	-2.697e-01	0.023988
##	2861	1.407e-02	-1.396e-01	-1.751e-01	0.175928
##	2871	3.163e-01	1.289e-01	6.000e-02	0.632319
##	2881	1.530e-01	-6.298e-01	1.045e-01	-0.171925
##	2891	2.034e-02	1.388e-01	3.488e-01	-0.063017
##	2901	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
##	2915	-3.976e-01	1.066e-01	-2.375e-01	-0.314632
##	2921	1.488e-01	-3.552e-01	-2.452e-01	-0.024711
##	2931	-9.659e-02	5.153e-02	-1.355e-01	0.226510
##	2941	2.047e-01	-1.382e-02	4.958e-02	0.435056
##	2951	1.369e-01	-2.076e-01	2.664e-01	0.228917
##	2961	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
##	2971	-1.499e-02	4.458e-02	-1.888e-01	-0.102870
##	2981	-4.086e-02	8.748e-03	1.588e-01	-0.298984
##	2991	2.130e-01	-1.837e-01	-4.465e-02	0.295367
##	3015	-4.680e-01	2.695e-01	-1.619e-01	0.012347
##	3021	-4.792e-01	1.136e-01	-1.841e-01	0.014748
##	3031	-1.892e-01	-1.676e-01	3.288e-01	-0.429939
##	3041	-1.663e-03	-3.996e-01	3.423e-01	-0.264864
##	3051	1.780e-01	1.963e-02	-2.343e-01	-0.631374
##	3061	2.671e-01	-1.096e-02	-7.822e-02	-0.502159
##	3071	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
##	3091	3.523e-01	-1.933e-01	-1.463e-01	0.196595
##	3105	2.385e-01	-1.712e-01	-2.070e-01	-0.508931
##	3115	2.196e-01	2.925e-01	-6.312e-03	-0.674355
##	3121	-2.747e-01	3.862e-02	2.041e-01	-0.261643
##	3131	-4.487e-01	5.393e-01	-1.234e-01	0.008192
##	3141	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
##	3151	-4.680e-01	2.695e-01	-1.619e-01	0.012347
##	3161	-3.765e-01	-2.693e-01	-3.217e-01	-0.501580
##	3171	2.465e-01	-1.791e-01	-1.169e-01	0.900758
##	3181	8.007e-02	-3.538e-01	-1.530e-01	0.611763
##	3191	2.655e-02	3.469e-02	-1.503e-01	0.173244
##	3201	-4.792e-01	1.136e-01	-1.841e-01	0.014748
##	3215	-1.304e-03	-1.233e-03	2.070e-01	0.267611
##	3221	-3.284e-01	-7.461e-02	-3.157e-01	-0.641327
##	3231	-1.016e-01	-5.725e-02	8.265e-01	-0.186959
##	3241	-2.872e-01	-1.357e-01	1.793e-01	-0.258959
##	3251	1.278e-01	-3.579e-01	-1.842e-01	0.419088
##	3261	-4.792e-01	1.136e-01	-1.841e-01	0.014748
##	3271	3.075e-01	-2.052e-01	1.534e-02	0.041807
##	3281	2.338e-01	-3.783e-01	-1.498e-01	0.100236
##	3291	1.995e-01	1.671e-01	-1.311e-01	0.274255
##	3301	-1.232e-01	-3.204e-01	-1.885e-01	0.232237
##	3315	-4.792e-01	1.136e-01	-1.841e-01	0.014748
##	3321	1.221e-01	-1.689e-01	-6.823e-02	-0.282804
##	3331	-1.832e-01	-5.030e-02	8.798e-01	0.142420
##	3341	-1.957e-01	-2.246e-01	8.550e-01	0.145104

## 3361	2.503e-01	1.404e-01	2.774e-02	-0.203844
## 3371	2.875e-01	-1.954e-01	-5.252e-02	-0.301269
## 3381	8.289e-02	-2.225e-01	4.487e-01	-0.740159
## 3391	3.381e-01	1.285e-01	2.050e-02	-0.030529
## 3401	3.461e-01	5.259e-01	1.715e-02	-0.214438
## 3415	2.912e-01	-1.433e-03	-1.798e-01	0.188963
## 3421	4.900e-01	-5.102e-02	6.195e-03	0.630736
## 3431	2.786e-01	-7.558e-03	-1.220e-01	-0.428720
## 3441	3.186e-01	4.198e-01	-8.067e-02	-0.051440
## 3451	1.840e-01	-1.662e-01	-1.730e-01	-0.217194
## 3461	2.089e-01	-1.729e-01	-1.560e-01	-0.685965
## 3471	1.156e-01	-1.866e-01	9.760e-02	-0.251759
## 3481	1.221e-01	-1.689e-01	-6.823e-02	-0.282804
## 3491	1.694e-01	1.010e-02	-1.288e-01	-0.182235
## 3501	2.630e-01	-3.518e-03	-1.389e-01	-0.506426
## 3515	9.422e-02	-1.562e-01	-1.249e-01	0.608720
## 3521	-2.098e-01	-4.222e-01	8.268e-01	0.148147
## 3531	2.863e-01	1.512e-01	-1.189e-01	-0.342701
## 3541	2.535e-01	5.531e-01	-9.293e-02	0.162240
## 3551	2.537e-01	2.663e-01	1.767e-01	-0.869891
## 3561	2.440e-01	9.295e-03	-2.205e-01	-0.183821
## 3571	2.938e-01	1.303e-01	5.647e-02	-0.365835
## 3581	1.346e-01	5.396e-03	-4.338e-02	-0.285488
## 3591	1.235e-01	-1.751e-01	-1.759e-02	-0.399631
## 3601	1.656e-01	-3.815e-01	-4.658e-02	-0.585615
## 3615	3.995e-01	1.401e-01	-1.543e-01	0.373619
## 3621	2.786e-01	-1.836e-01	-1.432e-01	-0.210326
## 3631	-4.579e-01	4.106e-01	-1.418e-01	0.010175
## 3641	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 3651	-2.098e-01	-4.222e-01	8.268e-01	0.148147
## 3661	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 3671	-1.116e-01	-5.740e-01	-3.887e-02	-0.084959
## 3681	-1.832e-01	-5.030e-02	8.798e-01	0.142420
## 3691	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 3701	-2.635e-01	1.945e-01	2.263e-01	-0.264044
## 3715	3.107e-01	-6.062e-03	-1.718e-01	-0.152516
## 3721	2.584e-01	1.350e-01	5.986e-02	-0.435956
## 3731	2.505e-01	-1.778e-01	-1.637e-01	-0.503742
## 3741	1.698e-01	-3.638e-01	-2.011e-01	-0.214151
## 3751	3.017e-01	-1.253e-02	-1.072e-01	-0.363926
## 3761	3.230e-01	2.951e-01	-1.499e-01	0.096661
## 3771	3.717e-01	-1.890e-01	-2.171e-01	-0.132821
## 3781	2.236e-01	-1.672e-01	-2.094e-01	0.086665
## 3791	3.510e-01	4.194e-01	-1.233e-01	-0.204939
## 3801	1.680e-01	-4.441e-02	3.479e-01	-0.569963
## 3815	2.593e-01	4.128e-01	6.062e-02	-0.013939
## 3821	3.107e-01	-6.062e-03	-1.718e-01	-0.152516
## 3831	8.248e-02	-1.482e-01	-1.910e-01	0.103618
## 3841	-5.751e-02	1.820e-01	6.892e-02	-0.096599
## 3851	-8.935e-02	-3.233e-01	-2.107e-01	0.095532
## 3861	-1.304e-03	-1.233e-03	2.070e-01	0.267611
## 3871	1.834e-01	2.871e-01	9.803e-02	-0.278173

## 3881	4.208e-01	1.375e-01	-1.695e-01	-0.088623
## 3891	4.294e-01	7.403e-01	-6.193e-02	-0.039576
## 3901	4.337e-01	5.129e-01	1.800e-02	-0.179131
## 3915	2.965e-01	-1.208e-02	-1.038e-01	-0.342803
## 3921	2.255e-01	-9.428e-03	-1.979e-02	0.292683
## 3931	3.375e-01	1.395e-01	-8.153e-02	-0.287220
## 3941	2.365e-01	2.990e-01	-7.983e-02	-0.363126
## 3951	2.667e-02	-1.664e-01	3.600e-02	-0.264194
## 3961	2.225e-01	1.689e-01	-1.862e-01	-0.124318
## 3971	2.687e-01	1.373e-01	2.089e-02	-0.675475
## 3981	2.874e-01	-1.802e-01	-1.827e-01	-0.099058
## 3991	4.071e-01	-2.003e-01	-1.618e-01	-0.023298
## 4001	2.047e-01	5.202e-01	2.468e-01	-0.585231
## 4015	1.061e-01	1.820e-01	-1.439e-01	0.098533
## 4021	6.516e-02	-3.564e-01	-1.256e-01	-0.102984
## 4031	2.065e-01	-1.678e-01	-1.889e-01	-0.253450
## 4041	3.266e-01	2.624e-01	1.238e-01	-0.466861
## 4051	1.694e-01	1.010e-02	-1.288e-01	-0.182235
## 4061	1.177e-01	1.213e-01	3.643e-01	-0.602372
## 4071	1.775e-01	1.548e-01	-8.375e-03	-0.416702
## 4081	1.411e-01	2.063e-02	-1.801e-01	0.032031
## 4091	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 4105	-1.957e-01	-2.246e-01	8.550e-01	0.145104
## 4115	1.471e-01	5.719e-01	-1.260e-01	-0.164515
## 4121	-1.232e-01	-3.204e-01	-1.885e-01	0.232237
## 4131	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 4151	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 4171	-2.872e-01	-1.357e-01	1.793e-01	-0.258959
## 4181	-2.793e-02	-3.731e-01	1.540e-01	0.273338
## 4191	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
## 4201	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 4215	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 4221	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 4231	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 4241	-3.013e-01	-3.333e-01	1.511e-01	-0.255916
## 4251	2.206e-01	-4.049e-01	9.804e-02	-0.142926
## 4261	2.173e-01	-3.390e-03	-7.322e-02	-0.210891
## 4271	3.358e-01	-1.421e-03	-2.498e-01	-0.318650
## 4281	1.930e-01	3.095e-03	-1.013e-01	-0.332783
## 4291	1.938e-01	-3.419e-03	-4.105e-02	-0.171707
## 4301	1.671e-01	-3.753e-01	-9.408e-02	-0.165980
## 4315	9.179e-02	1.551e-02	-7.262e-02	-0.108711
## 4321	3.775e-01	2.888e-01	-1.676e-01	0.024868
## 4331	2.138e-01	1.696e-01	-1.805e-01	-0.089103
## 4341	2.144e-01	5.609e-01	-1.146e-01	-0.038817
## 4351	4.173e-01	9.261e-01	1.033e-02	0.775014
## 4361	2.525e-01	-1.818e-01	-1.293e-01	-0.400882
## 4371	3.007e-01	-1.884e-01	-1.359e-01	-0.468905
## 4381	-1.097e-01	9.754e-01	1.026e+00	0.126625
## 4391	1.805e-01	7.953e-03	-1.306e-01	-0.466899
## 4401	1.774e-01	-1.829e-01	-1.759e-02	-0.295218
## 4415	2.417e-01	-1.688e-01	-2.213e-01	0.013414

## 4421	-5.221e-01	-4.864e-01	-2.697e-01	0.023988
## 4431	-1.957e-01	-2.246e-01	8.550e-01	0.145104
## 4441	-2.872e-01	-1.357e-01	1.793e-01	-0.258959
## 4451	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 4461	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
## 4471	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 4481	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 4491	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 4501	-2.872e-01	-1.357e-01	1.793e-01	-0.258959
## 4515	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 4521	3.403e-01	-1.946e-01	-1.234e-01	-0.021418
## 4531	2.667e-01	-1.755e-01	-2.043e-01	-0.455358
## 4541	2.258e-01	7.951e-03	-1.779e-01	0.125329
## 4551	2.048e-01	-1.777e-01	-9.807e-02	-0.208207
## 4561	5.556e-01	-9.923e-02	3.328e-01	-0.023274
## 4571	2.622e-01	1.562e-01	-1.325e-01	-0.412006
## 4581	3.164e-01	-6.551e-03	-1.756e-01	-0.175701
## 4591	3.139e-01	-1.907e-01	-1.286e-01	-0.282135
## 4601	3.167e-01	-1.925e-01	-1.192e-01	-0.393991
## 4615	2.018e-01	-1.964e-01	7.015e-02	-0.284690
## 4621	2.787e-01	-1.757e-01	-2.047e-01	0.191647
## 4631	3.453e-01	1.421e-01	-1.130e-01	-0.194936
## 4641	2.245e-01	1.584e-03	-1.240e-01	-0.077254
## 4651	5.721e-01	-2.710e-01	2.663e-01	0.861583
## 4661	-2.098e-01	-4.222e-01	8.268e-01	0.148147
## 4671	2.757e-01	-2.493e-02	4.913e-02	0.196978
## 4681	2.604e-01	-6.436e-03	-9.682e-02	0.094510
## 4691	2.574e-01	-1.701e-01	-2.315e-01	-0.050040
## 4701	1.624e-01	9.672e-01	-3.197e-02	0.086441
## 4715	2.912e-01	3.011e-01	-1.644e-01	-0.057296
## 4721	1.791e-01	-1.758e-01	-8.549e-02	-0.427567
## 4731	3.575e-01	-1.074e-02	-1.932e-01	-0.186778
## 4741	2.863e-01	5.133e-01	1.935e-01	-0.914610
## 4761	1.372e-01	4.438e-01	-3.930e-02	0.599480
## 4771	4.911e-01	2.652e-01	-1.029e-01	0.304804
## 4781	3.366e-01	8.265e-01	1.705e-01	0.183866
## 4791	4.014e-01	2.752e-01	-5.477e-02	1.130717
## 4801	2.271e-01	4.286e-01	-4.753e-02	-0.541132
## 4815	1.594e-01	4.207e-01	1.216e-01	-0.099239
## 4821	3.152e-01	5.401e-01	-7.483e-02	-0.534163
## 4831	2.111e-01	4.044e-01	2.026e-01	0.131340
## 4841	2.653e-01	2.977e-01	-1.015e-01	-0.133946
## 4851	-1.097e-01	9.754e-01	1.026e+00	0.126625
## 4861	1.141e-01	4.288e-01	1.154e-01	0.191284
## 4871	-4.978e-02	7.054e-01	-4.227e-02	0.216442
## 4881	-5.751e-02	1.820e-01	6.892e-02	-0.096599
## 4891	2.590e-01	-4.821e-03	-9.201e-02	0.898074
## 4901	-1.364e-01	6.035e-01	9.730e-01	0.132352
## 4915	1.972e-01	4.299e-01	-1.502e-02	-0.314417
## 4921	-1.720e-01	1.056e-01	9.021e-01	0.140019
## 4931	-1.282e-01	-4.291e-01	7.734e-01	-0.181232
## 4941	2.862e-01	1.499e-01	-9.930e-02	0.017316

## 4951	3.381e-01	-3.770e-03	-2.283e-01	-0.145717
## 4961	3.146e-01	1.357e-01	-2.006e-03	0.443390
## 4971	1.411e-01	2.063e-02	-1.801e-01	0.032031
## 4981	3.062e-01	4.008e-01	1.098e-01	0.190422
## 4991	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 5001	4.037e-01	-2.260e-02	-1.412e-01	0.179308
## 5015	1.357e-01	3.563e-03	-1.987e-02	0.118489
## 5021	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 5031	2.568e-01	-1.756e-01	-1.720e-01	0.433732
## 5041	4.079e-01	-2.062e-01	-1.058e-01	0.189216
## 5051	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 5061	2.847e-01	-2.175e-02	2.409e-03	-0.111818
## 5071	4.799e-01	-5.033e-02	7.663e-03	0.363046
## 5081	-1.832e-01	-5.030e-02	8.798e-01	0.142420
## 5091	-2.872e-01	-1.357e-01	1.793e-01	-0.258959
## 5105	2.573e-01	3.040e-01	-1.422e-01	0.079409
## 5115	1.964e-01	8.146e-03	-1.481e-01	-0.219878
## 5121	2.804e-01	1.546e-01	-1.249e-01	0.428647
## 5131	-1.527e-01	3.754e-01	9.405e-01	0.135865
## 5141	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 5151	-4.487e-01	5.393e-01	-1.234e-01	0.008192
## 5161	-8.116e-02	-1.483e-01	2.184e-02	-0.091514
## 5171	3.915e-02	7.931e-03	6.086e-02	-0.266878
## 5181	2.989e-01	-1.034e-02	-1.266e-01	-0.540271
## 5191	2.174e-01	-3.793e-01	-1.177e-01	0.189054
## 5201	-2.747e-01	3.862e-02	2.041e-01	-0.261643
## 5215	1.411e-01	2.063e-02	-1.801e-01	0.032031
## 5221	1.800e-01	5.648e-01	-1.025e-01	0.023651
## 5231	2.072e-01	-1.921e-01	2.715e-02	-0.157857
## 5241	3.046e-01	7.253e-01	2.446e-01	0.362051
## 5251	-5.763e-02	5.957e-01	-5.791e-02	0.218130
## 5261	4.551e-02	6.526e-01	3.002e-01	0.257543
## 5271	1.746e-01	5.477e-01	5.772e-02	0.110109
## 5281	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 5291	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 5301	8.357e-02	3.117e-03	5.468e-02	0.197840
## 5315	-8.543e-02	2.074e-01	-1.133e-01	0.224109
## 5321	2.410e-01	1.447e-01	-2.799e-03	-0.443475
## 5331	3.966e-02	4.086e-01	3.872e-01	-0.067172
## 5341	-4.402e-01	6.578e-01	-1.065e-01	0.006368
## 5351	-1.291e-01	7.056e-01	9.876e-01	0.130780
## 5361	-4.487e-01	5.393e-01	-1.234e-01	0.008192
## 5381	2.625e-01	-1.803e-01	-1.464e-01	0.022401
## 5391	3.935e-01	2.677e-01	1.074e-02	0.577772
## 5401	-4.487e-01	5.393e-01	-1.234e-01	0.008192
## 5415	-4.323e-01	7.674e-01	-9.088e-02	0.004680
## 5421	4.530e-01	2.763e-01	-1.504e-01	0.309235
## 5431	3.576e-01	-1.489e-02	-1.613e-01	-0.411667
## 5441	-7.533e-02	3.485e-01	-9.316e-02	0.221937
## 5461	2.022e-01	1.637e-01	-1.185e-01	-0.369992
## 5471	2.969e-01	2.720e-01	9.147e-02	0.192405
## 5481	-1.091e-01	-1.228e-01	-1.604e-01	0.229194

## 5491	3.948e-01	1.318e-01	-8.996e-02	-0.336689
## 5501	2.049e-01	1.525e-01	-1.882e-02	-0.174108
## 5515	2.590e-01	-4.821e-03	-9.201e-02	0.898074
## 5521	2.371e-01	5.190e-04	-1.445e-01	-0.697192
## 5531	-1.767e-01	6.704e-03	3.537e-01	-0.432623
## 5541	9.179e-02	1.551e-02	-7.262e-02	-0.108711
## 5551	1.808e-01	1.283e-01	2.295e-01	-0.077529
## 5561	2.245e-01	1.584e-03	-1.240e-01	-0.077254
## 5571	2.844e-01	-1.249e-02	-8.349e-02	-0.310652
## 5581	2.871e-01	1.308e-01	6.311e-02	-0.233939
## 5591	3.483e-01	1.232e-01	6.209e-02	0.317791
## 5601	6.897e-02	5.638e-01	5.046e-02	-0.091843
## 5615	2.653e-01	2.895e-01	-2.228e-02	0.178755
## 5621	1.698e-01	-3.638e-01	-2.011e-01	-0.214151
## 5631	1.091e-01	-2.436e-01	6.237e-01	0.323284
## 5641	1.280e-01	-1.225e-02	1.225e-01	-0.254443
## 5651	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 5661	2.130e-01	-1.837e-01	-4.465e-02	0.295367
## 5671	-1.832e-01	-5.030e-02	8.798e-01	0.142420
## 5681	1.232e-01	-1.707e-01	-4.473e-02	0.121173
## 5691	2.472e-01	1.630e-01	-1.623e-01	0.081580
## 5701	1.407e-02	-1.396e-01	-1.751e-01	0.175928
## 5715	-3.563e-02	9.030e-01	-1.410e-02	0.213399
## 5721	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 5731	2.655e-02	3.469e-02	-1.503e-01	0.173244
## 5741	2.785e-01	-3.989e-01	-1.526e-02	0.640447
## 5751	2.069e-01	-1.582e-02	5.649e-02	0.060424
## 5761	2.047e-01	-1.382e-02	4.958e-02	0.435056
## 5771	1.753e-02	-1.546e-01	-5.198e-02	-0.080779
## 5781	1.357e-01	1.248e-02	-1.046e-01	-0.144248
## 5791	-3.301e-03	-1.914e-01	3.017e-01	-0.057932
## 5801	2.255e-01	-9.428e-03	-1.979e-02	0.292683
## 5815	2.552e-01	1.567e-01	-1.217e-01	-0.131774
## 5821	3.146e-01	1.357e-01	-2.006e-03	0.443390
## 5831	-4.057e-01	1.139e+00	-3.785e-02	-0.001047
## 5841	-9.659e-02	5.153e-02	-1.355e-01	0.226510
## 5851	1.463e-01	2.288e-02	-2.118e-01	-0.192194
## 5861	-4.579e-01	4.106e-01	-1.418e-01	0.010175
## 5871	-9.788e-02	4.032e-02	-4.620e-02	-0.323580
## 5881	1.609e-01	-1.657e-01	-1.508e-01	-0.435406
## 5891	3.091e-01	-1.871e-01	-1.456e-01	0.112487
## 5901	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 5915	-1.619e-01	2.467e-01	9.222e-01	0.137848
## 5921	-5.751e-02	1.820e-01	6.892e-02	-0.096599
## 5931	1.544e-01	1.403e-02	-1.312e-01	0.413361
## 5941	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 5951	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 5961	1.067e-01	1.810e-02	-1.000e-01	0.606036
## 5971	-9.659e-02	5.153e-02	-1.355e-01	0.226510
## 5981	8.138e-02	-1.676e-01	-2.526e-02	-0.266502
## 5991	-1.720e-01	1.056e-01	9.021e-01	0.140019
## 6001	-1.291e-01	7.056e-01	9.876e-01	0.130780

## 6015	-9.659e-02	5.153e-02	-1.355e-01	0.226510
## 6021	-1.832e-01	-5.030e-02	8.798e-01	0.142420
## 6031	6.516e-02	-3.564e-01	-1.256e-01	-0.102984
## 6041	1.938e-01	-3.419e-03	-4.105e-02	-0.171707
## 6051	1.280e-01	-1.225e-02	1.225e-01	-0.254443
## 6061	-1.232e-01	-3.204e-01	-1.885e-01	0.232237
## 6071	1.386e-01	4.535e-01	-1.429e-01	-0.162691
## 6081	2.750e-01	-6.004e-03	-1.215e-01	0.019717
## 6091	-9.530e-02	-3.459e-01	-6.341e-03	-0.088472
## 6105	1.502e-01	2.919e-01	1.033e-01	-0.097257
## 6114	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 6121	2.395e-01	-4.039e-01	7.347e-02	0.315965
## 6131	1.232e-01	-1.707e-01	-4.473e-02	0.121173
## 6141	1.407e-02	-1.396e-01	-1.751e-01	0.175928
## 6151	-4.086e-02	8.748e-03	1.588e-01	-0.298984
## 6161	1.508e-01	-1.941e-01	1.128e-01	-0.488100
## 6171	4.117e-02	1.756e-01	-4.897e-03	-0.085864
## 6181	-1.957e-01	-2.246e-01	8.550e-01	0.145104
## 6191	-1.304e-03	-1.233e-03	2.070e-01	0.267611
## 6201	2.402e-01	3.035e-01	-1.217e-01	-0.260707
## 6214	-4.579e-01	4.106e-01	-1.418e-01	0.010175
## 6221	-1.378e-02	-1.755e-01	1.822e-01	0.270295
## 6231	2.718e-02	9.690e-03	5.536e-02	-0.535403
## 6241	-1.720e-01	1.056e-01	9.021e-01	0.140019
## 6251	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 6261	1.391e-01	-3.638e-01	-1.689e-01	-0.625483
## 6271	-1.619e-01	2.467e-01	9.222e-01	0.137848
## 6281	3.123e-01	-2.156e-01	1.197e-01	0.929151
## 6291	-1.832e-01	-5.030e-02	8.798e-01	0.142420
## 6301	-3.864e-01	2.626e-01	-2.152e-01	-0.317033
## 6314	-2.747e-01	3.862e-02	2.041e-01	-0.261643
## 6321	-3.645e-02	4.640e-02	-1.748e-01	-0.016231
## 6331	7.109e-02	-1.712e-01	2.983e-02	0.200524
## 6341	-1.655e-01	1.626e-01	3.759e-01	-0.435024
## 6351	6.221e-02	2.420e-02	-1.142e-01	-0.282836
## 6361	1.179e-01	1.740e-01	-7.777e-02	0.603635
## 6371	-7.533e-02	3.485e-01	-9.316e-02	0.221937
## 6381	1.866e-01	1.726e-01	-1.700e-01	-0.032839
## 6391	-1.378e-02	-1.755e-01	1.822e-01	0.270295
## 6401	1.758e-01	-1.632e-01	-1.782e-01	0.279340
## 6414	1.846e-01	-2.501e-02	1.798e-01	0.639618
## 6421	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 6431	-4.579e-01	4.106e-01	-1.418e-01	0.010175
## 6441	9.860e-03	1.547e-01	2.292e-01	0.265210
## 6451	-1.345e-01	6.249e-02	-1.899e-01	-0.122592
## 6461	1.249e-02	-2.020e-01	3.705e-01	-0.267907
## 6471	-1.957e-01	-2.246e-01	8.550e-01	0.145104
## 6481	1.091e-01	-2.436e-01	6.237e-01	0.323284
## 6491	-1.304e-03	-1.233e-03	2.070e-01	0.267611
## 6501	-8.543e-02	2.074e-01	-1.133e-01	0.224109
## 6513	3.579e-01	1.320e-01	-3.031e-02	0.268649
## 6521	2.146e-01	-5.251e-02	3.690e-01	-0.055722

## 6531	1.652e-01	-3.283e-02	2.560e-01	-0.238416
## 6551	2.559e-01	4.163e-01	4.091e-02	0.286127
## 6561	1.661e-01	4.293e-01	4.083e-02	0.111933
## 6571	3.692e-01	3.980e-01	5.162e-02	0.202894
## 6581	1.808e-01	1.283e-01	2.295e-01	-0.077529
## 6591	1.758e-01	-1.632e-01	-1.782e-01	0.279340
## 6601	2.400e-01	1.430e-01	2.215e-02	-0.009785
## 6613	1.067e-01	1.810e-02	-1.000e-01	0.606036
## 6621	4.028e-01	2.617e-01	3.396e-02	-0.222429
## 6631	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 6641	4.080e-02	-3.539e-02	4.393e-01	-0.422540
## 6651	-4.680e-01	2.695e-01	-1.619e-01	0.012347
## 6661	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 6671	1.694e-01	1.010e-02	-1.288e-01	-0.182235
## 6681	3.166e-01	-1.519e-02	-9.763e-02	-0.094518
## 6691	3.652e-01	-2.434e-02	-6.973e-02	0.420544
## 6701	3.044e-02	2.799e-01	3.689e-01	-0.065189
## 6713	-8.116e-02	-1.483e-01	2.184e-02	-0.091514
## 6721	-5.058e-01	-2.583e-01	-2.371e-01	0.020475
## 6731	2.047e-01	-1.382e-02	4.958e-02	0.435056
## 6741	-8.543e-02	2.074e-01	-1.133e-01	0.224109
## 6751	1.789e-01	-1.800e-01	-4.988e-02	-0.512950
## 6761	-1.091e-01	-1.228e-01	-1.604e-01	0.229194
## 6771	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 6781	8.248e-02	-1.482e-01	-1.910e-01	0.103618
## 6791	2.912e-01	-1.433e-03	-1.798e-01	0.188963
## 6801	-4.792e-01	1.136e-01	-1.841e-01	0.014748
## 6813	1.270e-01	1.624e-01	-1.088e-02	-0.513672
## 6821	-7.533e-02	3.485e-01	-9.316e-02	0.221937
## 6831	-4.916e-01	-6.070e-02	-2.090e-01	0.017432
## 6841	4.803e-01	-2.196e-01	-8.160e-02	0.232821
## 6851	-5.751e-02	1.820e-01	6.892e-02	-0.096599
## 6861	-2.098e-01	-4.222e-01	8.268e-01	0.148147
## 6871	1.857e-01	1.584e-02	-1.989e-01	-0.071785
## 6881	1.606e-01	1.226e-01	3.135e-01	0.223832
## 6891	1.407e-02	-1.396e-01	-1.751e-01	0.175928
## 6901	1.488e-01	-3.552e-01	-2.452e-01	-0.024711
## 6913	3.292e-01	-1.194e-02	-1.402e-01	0.034333
## 6921	-4.250e-01	8.695e-01	-7.632e-02	0.003107

```

# have the scores reversed to have the adult lifespan to the left
test.pca_WA1_reverse <- test.pca_WA1
test.pca_WA1_reverse$CA$u <- -(test.pca_WA1_reverse$CA$u)
test.pca_WA1_reverse$CA$v <- -(test.pca_WA1_reverse$CA$v)
g <- vegan::scores(test.pca_WA1_reverse, display = "species", scaling=2)

# Figure
par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
plot(test.pca_WA1_reverse, scaling = scl, choices=c(1,2), type="n", ylim=c(-4,3), xlim=c(-3.5, 3.5), main="", xlab="PC1 = 52%", ylab="PC2 = 25%", cex.axis=2, cex.lab=2)
with(data_WA, points(test.pca_WA1_reverse, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_WA1_reverse, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1]+0.5, y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1]-0.75, y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)

```



```
# Test for proportion of variance explained per axe
sum_all_WA <- summary(test.pca_WA1)

prop_WA <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_WA[j,i+1] <- as.numeric(abs(sum_all_WA$species[i,j])/sum(abs(sum_all_WA$species[i,])))
  }
}

prop_WA
```

```
##      PC      prim      adult      breed      freq
## 1  1 0.118024472 0.62816690 0.61018610 0.37558789
## 2  2 0.790849517 0.02568004 0.01324922 0.09977922
## 3  3 0.089221100 0.15587848 0.19373423 0.51247499
## 4  4 0.001904911 0.19027458 0.18283045 0.01215791
```

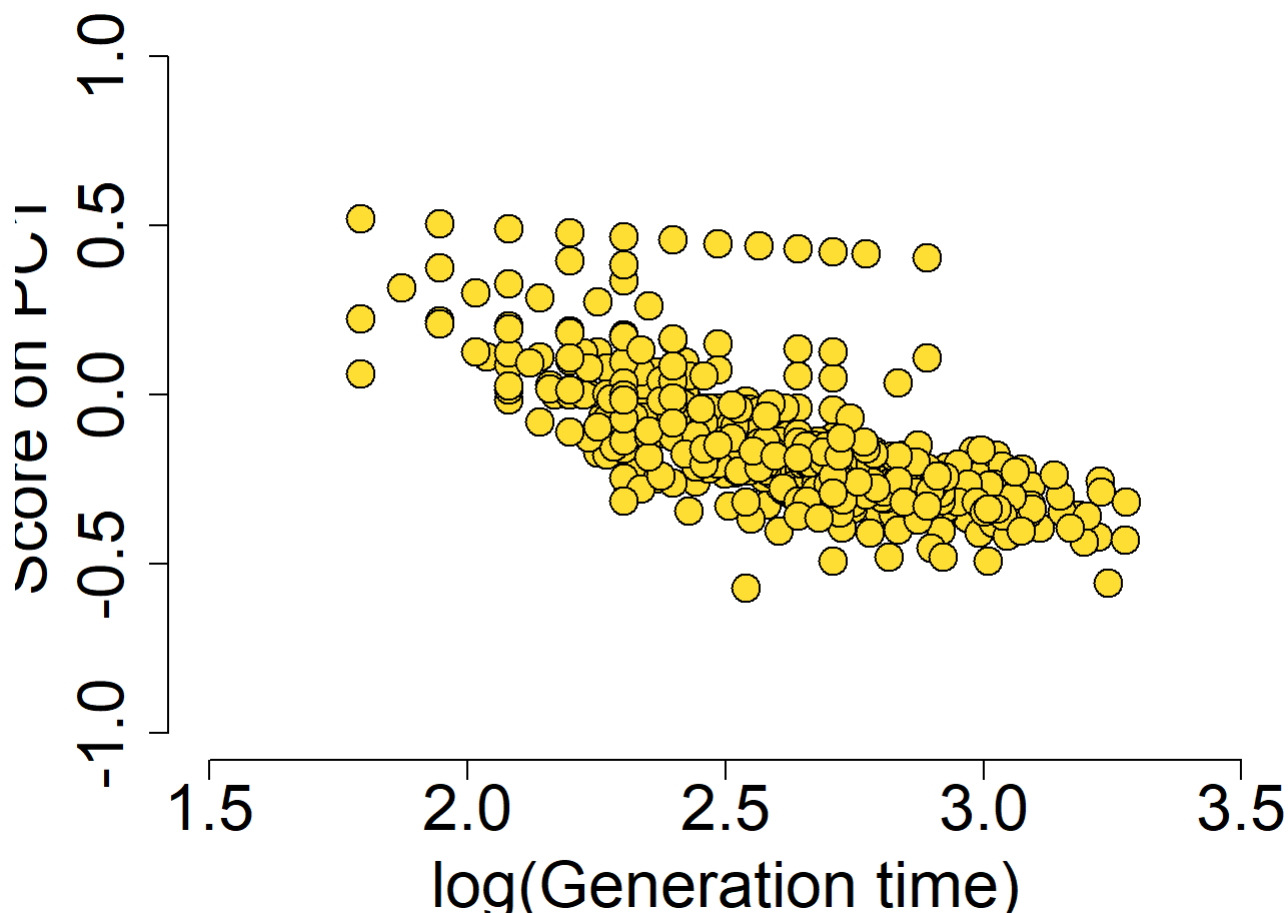
```
# Covariances
test <- as.matrix(scale(log(data_WA[,c(2,3,4,7)])))
cov(test)
```

```
##           age_first_repro lifetime_adult breed_freq fec_one
## age_first_repro      1.00000000      0.07416693 -0.08194516 -0.02148279
## lifetime_adult      0.07416693      1.00000000 -0.84314440 -0.35990610
## breed_freq        -0.08194516     -0.84314440      1.00000000  0.31150143
## fec_one           -0.02148279     -0.35990610  0.31150143      1.00000000
```

```
# Correlation with generation time
individual_score_WA_reverse <- as.numeric(vegan::scores(test.pca_WA1_reverse)$sites[,1])
summary(lm(individual_score_WA_reverse ~ log(data_WA$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_WA_reverse ~ log(data_WA$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.52226 -0.12872 -0.03793  0.08400  0.68963
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.64152     0.05544   29.61  <2e-16 ***
## log(data_WA$generation_time) -0.66615     0.02232  -29.85  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1823 on 678 degrees of freedom
## Multiple R-squared:  0.5679, Adjusted R-squared:  0.5672
## F-statistic: 891 on 1 and 678 DF, p-value: < 2.2e-16
```

```
par(bty="n")
plot(log(data_WA$generation_time), individual_score_WA_reverse, pch=21, xlab="log(Generation time)", ylab="Score on PC1", xlim=c(1.5,3.5), ylim=c(-1,1), bg=colors[15], cex=2, cex.axis=2, cex.lab=2)
```



Black-browed albatross

```
data_bbalbatross <- data[which(data$species == "blackbrowedalbatross"),]  
data_bbalbatross$fec_one <- data_bbalbatross$fecundity + 1  
test.pca_bbalbatross1 <- rda(log(data_bbalbatross[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_bbalbatross1)
```

```
##
## Call:
## rda(X = log(data_bbalbatross[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained    4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      1.5869 1.1718 0.8787 0.36265
## Proportion Explained 0.3967 0.2929 0.2197 0.09066
## Cumulative Proportion 0.3967 0.6897 0.9093 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 5.208581
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro 0.5545 1.7262 -1.8582 -0.2048
## lifetime_adult  2.3146 -0.1954 0.5275 -1.0530
## breed_freq      -2.1204 0.9693 0.5914 -0.9985
## fec_one         0.7761 1.9975 1.3705 0.5585
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 1337  0.475739 0.049126 0.361610 -0.201288
## 2130 -0.135435 1.183096 -0.880124 0.258500
## 3127  0.556709 -0.281721 -0.445510 -0.273977
## 4128 -0.115454 0.546472 0.075002 0.163823
## 5126  0.305474 0.085496 -0.150722 -0.533831
## 6126  0.627702 -0.173092 -0.116525 0.127833
## 7122  0.641591 -0.346867 -0.348918 0.809397
## 8122 -0.438919 0.048658 -0.878795 -0.197458
## 9120 -0.121239 0.138539 0.256704 -0.112251
## 10120 -0.531464 -0.341482 -0.318721 -0.047912
## 11118 0.774419 -0.594753 -0.192934 0.212005
## 12107 -0.364984 -0.360512 -0.250196 -0.379320
## 1338 -0.359421 0.383800 -1.359916 -0.325921
## 1429  0.023502 0.201378 0.377536 -0.287043
## 1530  0.187539 -0.378380 0.186516 -0.107979
## 1629  0.113270 0.223353 0.439360 -0.419717
## 1730  0.003684 0.627048 -0.016566 -0.065191
```

##	1830	0.472828	-0.001308	0.029787	-0.171179
##	1928	0.263959	0.282570	0.560369	-0.610553
##	2028	0.124362	0.414931	0.329148	-0.322586
##	2136	1.137365	-0.148428	-0.672209	0.535114
##	2228	0.446018	0.552031	-0.113155	-0.505482
##	2328	0.072124	0.232839	0.162534	-0.487100
##	2427	0.684620	0.020037	-0.341018	-0.393882
##	2527	0.322428	0.127039	0.125585	-0.365848
##	2625	0.348332	0.235249	0.211277	-0.161122
##	2725	-0.240320	0.740933	-0.245367	0.427987
##	2825	-0.181904	-0.095174	0.344870	-0.158070
##	2925	-0.240320	0.740933	-0.245367	0.427987
##	3025	0.677214	0.011856	0.111294	-0.246349
##	3128	0.455644	0.033099	-0.140925	-0.657753
##	3225	0.491377	0.173535	-0.227992	-0.648881
##	3325	-0.221928	0.157774	-0.042359	-0.056900
##	3425	-0.332864	0.350793	0.314707	0.577532
##	3525	0.562013	-0.172472	-0.825977	0.169730
##	3624	0.226338	-0.004104	0.619187	-0.774100
##	3724	0.271930	0.639990	-0.246345	-0.758352
##	3825	0.300792	-0.117276	0.219656	-0.789597
##	3924	0.570342	-0.414907	0.230619	0.220524
##	4025	-0.115454	0.546472	0.075002	0.163823
##	4129	-0.281933	0.565503	0.006476	0.495231
##	4225	0.328564	0.166343	0.148228	-0.223376
##	4325	0.041813	0.787788	-0.247320	-0.126804
##	4425	-0.504581	-0.228151	-0.481416	-0.091354
##	4525	0.316473	0.147320	-0.122352	-0.240345
##	4624	0.324448	0.174156	-0.330182	-0.168403
##	4724	0.362221	0.061475	-0.021117	0.520443
##	4824	0.360191	0.282866	-0.015147	0.047820
##	4924	0.241206	-0.228973	0.001930	0.083182
##	5024	0.072549	0.214698	0.412328	-0.357658
##	5127	-0.332864	0.350793	0.314707	0.577532
##	5224	-0.221928	0.157774	-0.042359	-0.056900
##	5324	0.401149	-0.194801	0.162891	-0.656232
##	5424	0.568830	0.418618	-0.226651	-0.267628
##	5524	0.573128	0.245238	-0.417694	0.016061
##	5624	0.600272	-0.039109	-0.703950	0.290218
##	5724	0.386452	-0.322018	0.002733	-0.315096
##	5824	0.054380	0.569509	-0.230343	-0.355318
##	5924	-0.338101	-0.247181	-0.412891	-0.422761
##	6024	0.391053	0.249427	0.280090	-0.550345
##	6127	0.452184	-0.603237	-0.569468	0.136739
##	6223	-0.458779	-0.035065	-0.758605	-0.165366
##	6323	-0.042117	0.433962	0.260623	0.008822
##	6423	0.090336	0.143411	0.366213	-0.491942
##	6520	0.275777	-0.055199	0.225232	-0.384660
##	6620	-0.176126	0.350860	-0.319548	-0.130912
##	6720	-0.638738	-0.793721	0.330501	0.125436
##	6820	0.322428	0.127039	0.125585	-0.365848
##	6920	0.573662	-0.240179	-0.169203	-0.105994

##	7019	-0.221928	0.157774	-0.042359	-0.056900
##	7123	-0.070308	0.353248	-0.051527	-0.194551
##	7219	-0.082331	0.025412	0.188862	-0.344867
##	7319	0.245299	-0.183683	0.409680	-0.335411
##	7419	-0.305981	0.464125	0.152012	0.534091
##	7519	-0.332864	0.350793	0.314707	0.577532
##	7619	-0.281933	0.565503	0.006476	0.495231
##	7719	-0.123149	-0.332061	0.191452	-0.603655
##	7819	-0.082331	0.025412	0.188862	-0.344867
##	7919	-0.332864	0.350793	0.314707	0.577532
##	8019	-0.115454	0.546472	0.075002	0.163823
##	8123	-0.139501	0.445094	0.220538	0.202683
##	8218	0.586701	-0.316013	-0.247296	-0.001291
##	8319	0.335338	-0.051856	0.141578	0.563884
##	8419	0.235064	-0.195874	0.370327	-0.108794
##	8518	-0.281933	0.565503	0.006476	0.495231
##	8619	0.669431	-0.458287	-0.129525	0.502647
##	8718	-0.364984	-0.360512	-0.250196	-0.379320
##	8819	-0.363341	0.222309	0.499156	0.626782
##	8919	-0.093700	0.638180	-0.056651	0.128671
##	9019	-0.112809	-0.103072	0.373310	-0.295617
##	9122	-0.221928	0.157774	-0.042359	-0.056900
##	9218	0.279465	-0.095609	0.123957	0.203671
##	9318	-0.480533	-0.126772	-0.626952	-0.130213
##	9417	-0.480533	-0.126772	-0.626952	-0.130213
##	9517	-0.638738	-0.793721	0.330501	0.125436
##	9617	0.508180	-0.421546	0.241533	-0.102805
##	9718	-0.561941	-0.469966	-0.134273	0.001337
##	9818	-0.166384	0.331763	0.383233	0.246124
##	9918	-0.363341	0.222309	0.499156	0.626782
##	10018	0.374092	-0.327747	0.007478	-0.439920
##	10122	0.281924	-0.140797	-0.145389	-0.274372
##	10218	-0.638738	-0.793721	0.330501	0.125436
##	10318	0.368131	-0.296887	-0.375097	0.727730
##	10418	-0.248811	0.044443	0.120336	-0.013459
##	10518	-0.561941	-0.469966	-0.134273	0.001337
##	10618	-0.504581	-0.228151	-0.481416	-0.091354
##	10718	-0.174658	1.017742	-0.642746	0.321883
##	10818	0.434569	-0.134672	-0.092239	-0.291667
##	10917	-0.232045	0.054954	0.780612	0.352228
##	11017	-0.078982	-0.178105	0.332322	-0.464561
##	11119	-0.279288	-0.084042	0.304785	0.035791
##	11217	0.374186	0.114095	-0.166249	0.034198
##	11316	0.341248	-0.410218	-0.212401	0.771171
##	11417	-0.139501	0.445094	0.220538	0.202683
##	11517	-0.279288	-0.084042	0.304785	0.035791
##	11617	0.260693	-0.334297	0.185157	0.474039
##	11717	0.328564	0.166343	0.148228	-0.223376
##	11817	0.309017	-0.015322	0.586781	-0.350712
##	11917	0.461941	-0.317275	-0.532614	-0.039915
##	12017	-0.121239	0.138539	0.256704	-0.112251
##	12117	-0.597125	-0.618290	0.078658	0.058192

##	12217	-0.364984	-0.360512	-0.250196	-0.379320
##	12317	0.392179	-0.195509	-0.520632	0.688871
##	12416	-0.504581	-0.228151	-0.481416	-0.091354
##	12516	0.549047	-0.737006	0.211156	0.958942
##	12616	0.006489	0.677002	-0.516300	-0.318650
##	12716	-0.597125	-0.618290	0.078658	0.058192
##	12816	0.304861	-0.180341	0.326027	0.613134
##	12914	0.082422	-0.047564	0.858685	-0.368823
##	13011	0.261111	-0.179145	0.774531	-0.230468
##	13113	0.379096	-0.691158	0.280632	0.254768
##	13211	-0.561941	-0.469966	-0.134273	0.001337
##	1339	-0.158291	0.029007	-0.109909	-0.357040
##	1347	0.414923	-0.421710	0.539759	0.271711
##	1357	0.374706	-0.315839	0.175966	1.021769
##	1367	0.309260	0.068157	0.293084	0.130120
##	1377	-0.561941	-0.469966	-0.134273	0.001337
##	1387	0.145233	0.521115	-0.027556	-0.475111
##	1397	-0.480533	-0.126772	-0.626952	-0.130213
##	1407	-0.363341	0.222309	0.499156	0.626782
##	14113	-0.531464	-0.341482	-0.318721	-0.047912
##	14210	-0.504581	-0.228151	-0.481416	-0.091354
##	1437	0.515996	0.279802	-0.679121	0.793453
##	1447	0.114096	-0.660068	0.277419	-0.171613
##	1457	0.405183	-0.187355	-0.008483	0.972520
##	1467	0.038097	-0.038681	0.199600	-0.656455
##	1477	0.012767	0.394078	0.021500	-0.288073
##	1487	0.318054	-0.092482	-0.161987	0.381349
##	1497	-0.531464	-0.341482	-0.318721	-0.047912
##	1507	-0.093700	0.638180	-0.056651	0.128671
##	15113	-0.314472	-0.232366	0.517715	0.092645
##	15210	-0.186900	-0.138270	0.654083	-0.006147
##	1537	-0.065566	0.035924	0.849137	0.020820
##	1547	-0.561941	-0.469966	-0.134273	0.001337
##	1567	-0.597125	-0.618290	0.078658	0.058192
##	1577	0.174798	-0.411521	0.187888	-0.434292
##	1587	0.670243	-0.839821	-0.185877	0.688627
##	1596	-0.504581	-0.228151	-0.481416	-0.091354
##	1607	0.497727	0.202785	-0.568557	0.822974
##	16113	-0.438919	0.048658	-0.878795	-0.197458
##	16210	-0.015425	-0.114204	0.413395	-0.489478
##	1636	-0.093700	0.638180	-0.056651	0.128671
##	1647	-0.561941	-0.469966	-0.134273	0.001337
##	1657	-0.597125	-0.618290	0.078658	0.058192
##	1667	-0.115454	0.546472	0.075002	0.163823
##	1677	0.285705	-0.584207	-0.637993	0.468147
##	1687	0.026978	0.426064	0.289063	-0.128725
##	1697	-0.398525	0.073985	0.712086	0.683636
##	1707	-0.281933	0.565503	0.006476	0.495231
##	17113	0.026978	0.426064	0.289063	-0.128725
##	17210	-0.314053	-0.145803	-0.558427	-0.461621
##	1736	-0.248811	0.044443	0.120336	-0.013459
##	1747	0.152356	-0.526705	0.399446	-0.051125

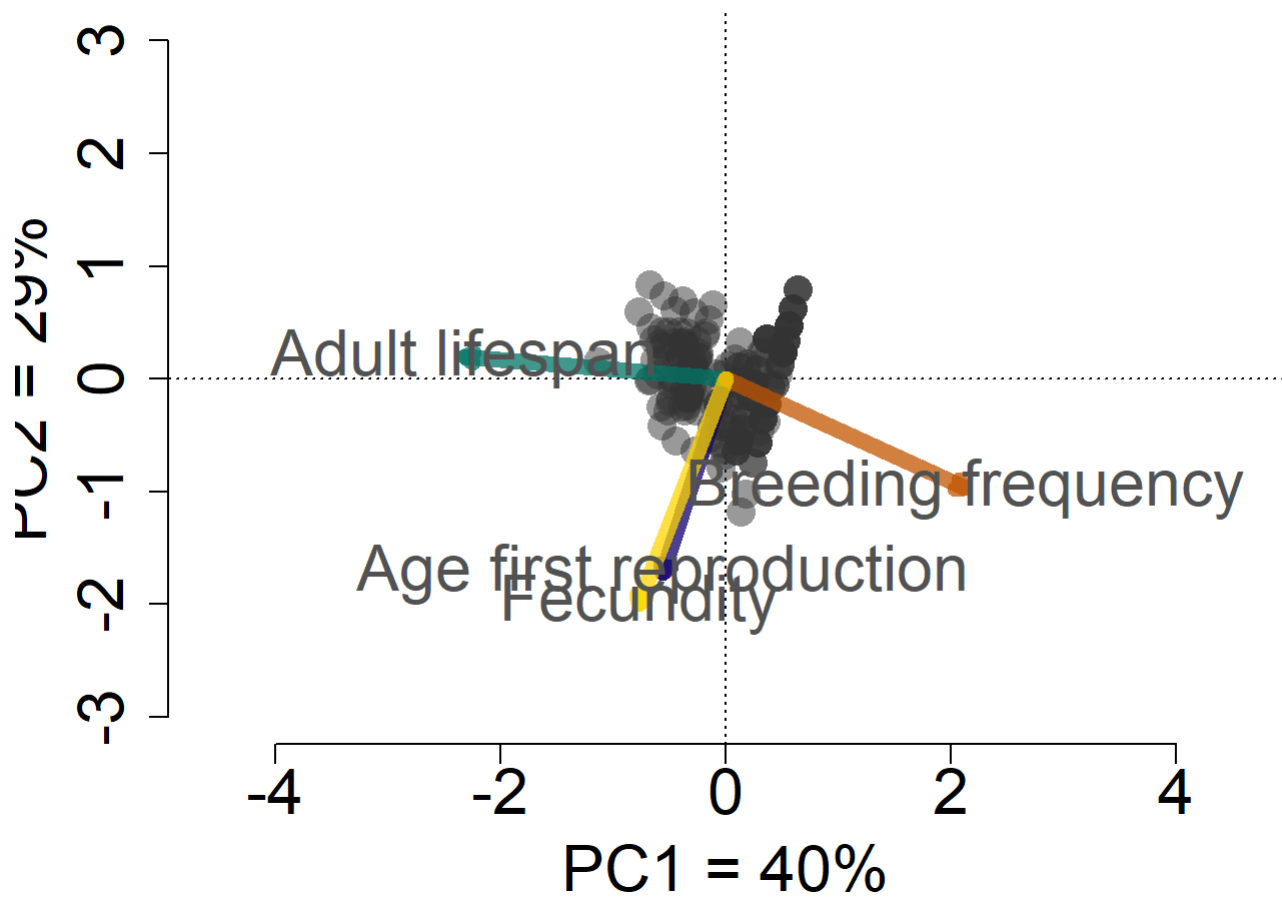
```
## 1757 -0.398525 0.073985 0.712086 0.683636
## 1767 -0.248811 0.044443 0.120336 -0.013459
## 1777 -0.115454 0.546472 0.075002 0.163823
## 1787 -0.112809 -0.103072 0.373310 -0.295617
## 1797 -0.364984 -0.360512 -0.250196 -0.379320
## 1806 -0.166384 0.331763 0.383233 0.246124
## 18113 -0.504581 -0.228151 -0.481416 -0.091354
## 18210 -0.030382 0.184248 0.636207 -0.036034
## 1837 -0.504581 -0.228151 -0.481416 -0.091354
## 1847 -0.531464 -0.341482 -0.318721 -0.047912
## 1857 -0.364984 -0.360512 -0.250196 -0.379320
## 1867 -0.363341 0.222309 0.499156 0.626782
```

have the scores reversed to have the adult lifespan to the left

```
test.pca_bbalbatross1_reverse <- test.pca_bbalbatross1
test.pca_bbalbatross1_reverse$CA$u <- -(test.pca_bbalbatross1_reverse$CA$u)
test.pca_bbalbatross1_reverse$CA$v <- -(test.pca_bbalbatross1_reverse$CA$v)
g <- vegan::scores(test.pca_bbalbatross1_reverse, display = "species", scaling=2)
```

Figure

```
par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
plot(test.pca_bbalbatross1_reverse, scaling = scl, choices=c(1,2), type="n", ylim=c(-3,3), xlim=c(-3.5,3.5), main="", xlab="PC1 = 40%", ylab="PC2 = 29%", cex.lab=2, cex.axis=2)
with(data_bbalbatross, points(test.pca_bbalbatross1_reverse, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_bbalbatross1_reverse, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```
# Test for proportion of variance explained per axe
sum_all_bbalbatross <- summary(test.pca_bbalbatross1)

prop_bbalbatross <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_bbalbatross[j,i+1] <- as.numeric(abs(sum_all_bbalbatross$species[i,j])/sum(abs(sum_all_bbalbatross$species[i,])))
  }
}

prop_bbalbatross
```

```
##      PC      prim      adult      breed      freq
## 1  1 0.12766388 0.56585337 0.4531121 0.1650286
## 2  2 0.39739817 0.04776178 0.2071259 0.4247620
## 3  3 0.42779351 0.12896306 0.1263846 0.2914390
## 4  4 0.04714444 0.25742179 0.2133774 0.1187703
```

Covariances

```
test <- as.matrix(scale(log(data_bbalbatross[,c(2,3,4,7)])))
cov(test)
```

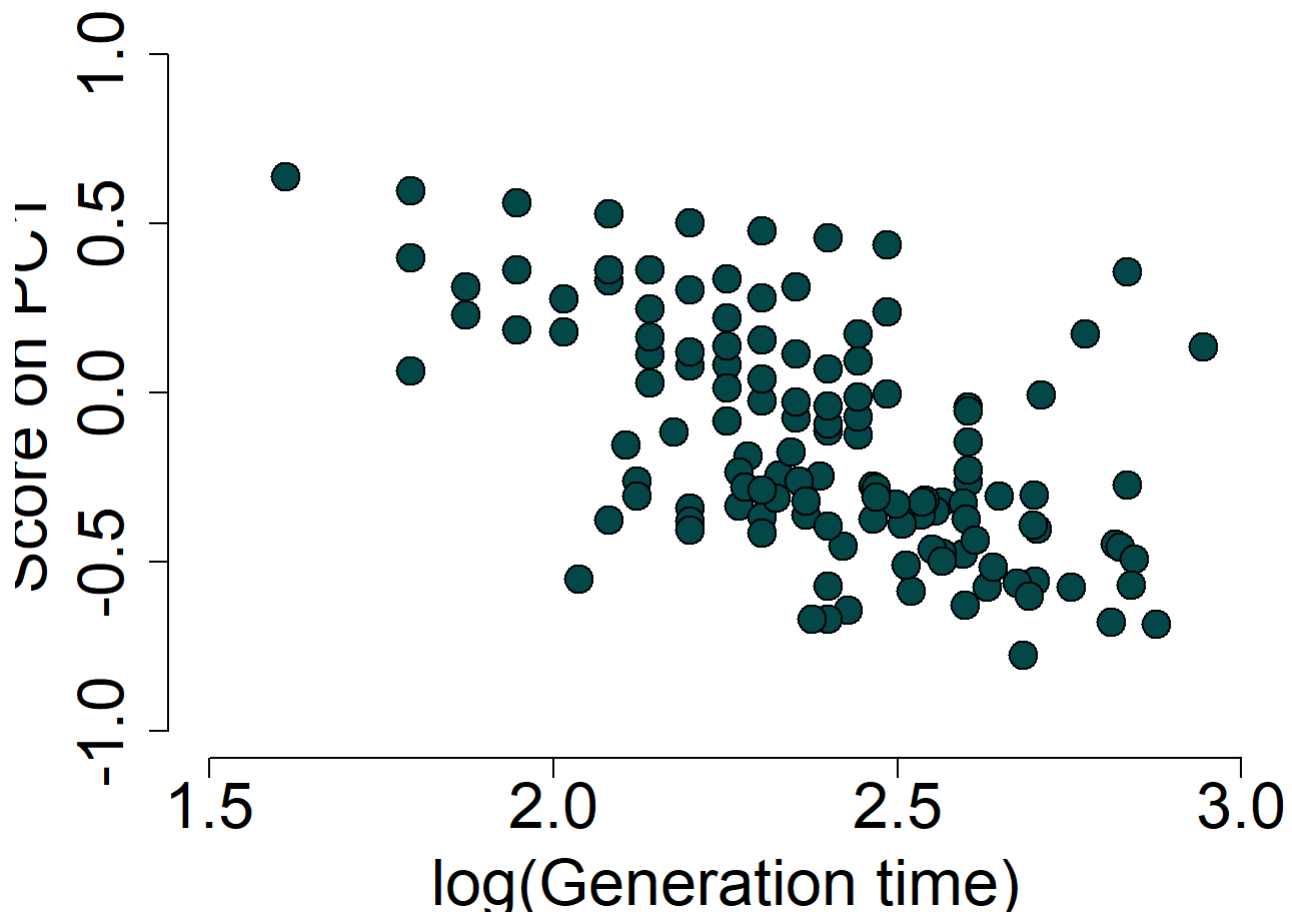
```
##           age_first_repro lifetime_adult breed_freq fec_one
## age_first_repro      1.00000000      0.02678666 -0.05856752 0.17947962
## lifetime_adult      0.02678666      1.00000000 -0.55050839 0.22718498
## breed_freq          -0.05856752     -0.55050839  1.00000000 0.08011988
## fec_one             0.17947962      0.22718498  0.08011988 1.00000000
```

Correlation with generation time

```
individual_score_bbalbatross_reverse <- as.numeric(vegan::scores(test.pca_bbalbatross1_reverse)
$sites[,1])
summary(lm(individual_score_bbalbatross_reverse ~ log(data_bbalbatross$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_bbalbatross_reverse ~ log(data_bbalbatross$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.81929 -0.17548  0.00461  0.16026  0.86381
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      2.25161    0.17371   12.96  <2e-16 ***
## log(data_bbalbatross$generation_time) -0.97275    0.07453  -13.05  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2771 on 183 degrees of freedom
## Multiple R-squared:  0.4821, Adjusted R-squared:  0.4793
## F-statistic: 170.4 on 1 and 183 DF, p-value: < 2.2e-16
```

```
par(bty="n")
plot(log(data_bbalbatross$generation_time), individual_score_bbalbatross_reverse, pch=21, xlab
="log(Generation time)", ylab="Score on PC1", xlim=c(1.5,3), ylim=c(-1,1), bg=colors[3], cex=2,
cex.axis=2, cex.lab=2)
```



Snow petrel

```
data_petrel <- data[which(data$species == "snowpetrel"),]  
data_petrel$fec_one <- data_petrel$fecundity + 1  
test.pca_petrel1 <- rda(log(data_petrel[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_petrel1)
```

```
##
## Call:
## rda(X = log(data_petrel[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained    4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1      PC2      PC3      PC4
## Eigenvalue      1.6489 1.1039 0.8881 0.35908
## Proportion Explained 0.4122 0.2760 0.2220 0.08977
## Cumulative Proportion 0.4122 0.6882 0.9102 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 3.439791
##
##
## Species scores
##
##           PC1      PC2      PC3      PC4
## age_first_repro -0.5948 0.928424 -1.31979 0.02154
## lifetime_adult  1.5170 -0.006964 -0.40059 -0.70434
## breed_freq      -1.4859 -0.498022 0.05028 -0.70689
## fec_one         0.1207 -1.468147 -0.84976 0.25675
##
##
## Site scores (weighted sums of species scores)
##
##           PC1      PC2      PC3      PC4
## 1305  0.47036 0.10942 -0.9754955 -0.563479
## 4107  0.95685 -0.04646 0.1256678 -0.277999
## 6107  0.90286 -0.13051 0.0923567 -0.187150
## 7106 -0.70602 0.66319 0.0893757 0.005192
## 8106 -0.07034 0.18213 -1.1321806 0.790148
## 10106 0.36859 -0.29181 0.2223977 -0.020253
## 11106 -0.65786 0.55089 0.2878182 -0.002819
## 12102 0.73757 0.84253 -0.4234808 0.713841
## 1319 -0.57108 0.76103 -0.1726555 -0.366030
## 1519 0.82179 0.74976 -0.5185788 1.210971
## 1719 -0.58994 -0.68300 -0.5999619 0.660580
## 1819 -0.74851 0.76225 -0.0856630 0.012258
## 1918 -0.08650 0.17649 0.7424402 -0.907679
## 2018 0.31847 0.03433 0.0275879 0.159503
## 2118 0.25623 -0.05039 -0.2150545 -0.825765
## 2218 0.34672 0.10542 0.5081990 -0.233820
## 2318 -0.72789 0.71418 -0.0007192 0.008829
```

```
## 2418 -0.16721 -0.73195 0.1522873 -0.139993
## 2518 0.25341 -0.25815 -0.3541113 -0.468905
## 2617 -0.53434 -0.81264 -0.3708769 0.651333
## 2717 0.01022 -0.73316 0.0652948 -0.518280
## 2817 -0.53649 0.26792 0.7878532 -0.023004
## 2917 -0.33992 -1.26593 0.4301317 0.618999
## 3017 -0.74851 0.76225 -0.0856630 0.012258
## 3117 0.15265 0.25251 0.6905971 -0.757012
## 3217 0.18541 -0.86837 0.1508598 -0.785177
## 3317 -0.76801 0.80772 -0.1660129 0.015501
## 3417 -0.50295 -0.88581 -0.2415683 0.646113
## 3517 -0.31933 -0.45507 0.1815426 -0.013227
## 3817 0.69071 0.18037 1.3297793 0.692970
## 4017 0.36864 -0.25029 -0.9078535 -0.910999
## 4117 -0.40784 -0.03204 1.3179118 -0.044400
## 4217 1.33393 0.23518 0.7400894 1.129493
## 4317 0.25532 0.16561 -0.8312666 -0.086248
## 4417 -0.50295 -0.88581 -0.2415683 0.646113
## 4517 0.55594 0.05821 -0.6194796 -0.841863
```

```
# have the scores reversed to have the adult lifespan to the left
```

```
test.pca_petrel1_reverse <- test.pca_petrel1
```

```
test.pca_petrel1_reverse$CA$u <- -(test.pca_petrel1_reverse$CA$u)
```

```
test.pca_petrel1_reverse$CA$v <- -(test.pca_petrel1_reverse$CA$v)
```

```
g <- vegan::scores(test.pca_petrel1_reverse, display = "species", scaling=2)
```

```
# Figure
```

```
par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
```

```
plot(test.pca_petrel1_reverse, scaling = scl, choices=c(1,2), type="n", ylim=c(-3,3), xlim=c(-3.5,3.5), main="", xlab="PC1 = 41%", ylab="PC2 = 28%")
```

```
with(data_petrel, points(test.pca_petrel1_reverse, display = "sites", col = col_points, scaling = scl, pch = 16, cex=1.5))
```

```
g <- vegan::scores(test.pca_petrel1_reverse, display = "species")
```

```
len <- 1
```

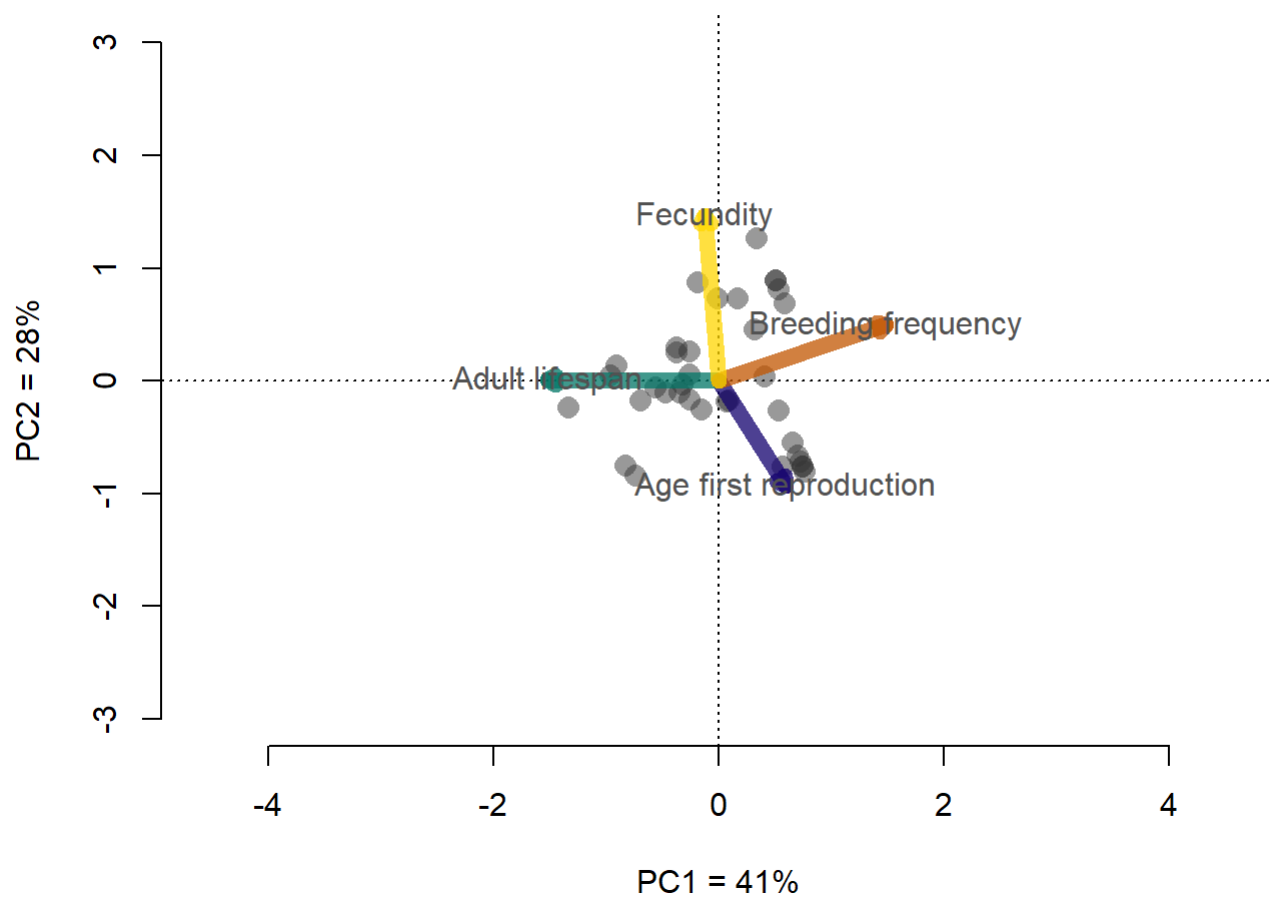
```
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
```

```
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32")
```

```
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32")
```

```
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32")
```

```
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32")
```



```
# Test for proportion of variance explained per axe
sum_all_petrel <- summary(test.pca_petrel1)

prop_petrel <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_petrel[j,i+1] <- as.numeric(abs(sum_all_petrel$species[i,j])/sum(abs(sum_all_petrel$species[i,])))
  }
}

prop_petrel
```

```
##      PC      prim      adult      breed      freq
## 1  1 0.207637361 0.577055924 0.54207781 0.04478333
## 2  2 0.324109532 0.002648911 0.18168977 0.54469307
## 3  3 0.460733021 0.152375768 0.01834335 0.31526669
## 4  4 0.007520085 0.267919396 0.25788906 0.09525691
```

Covariances

```
test <- as.matrix(scale(log(data_petrel[,c(2,3,4,7)])))
cov(test)
```

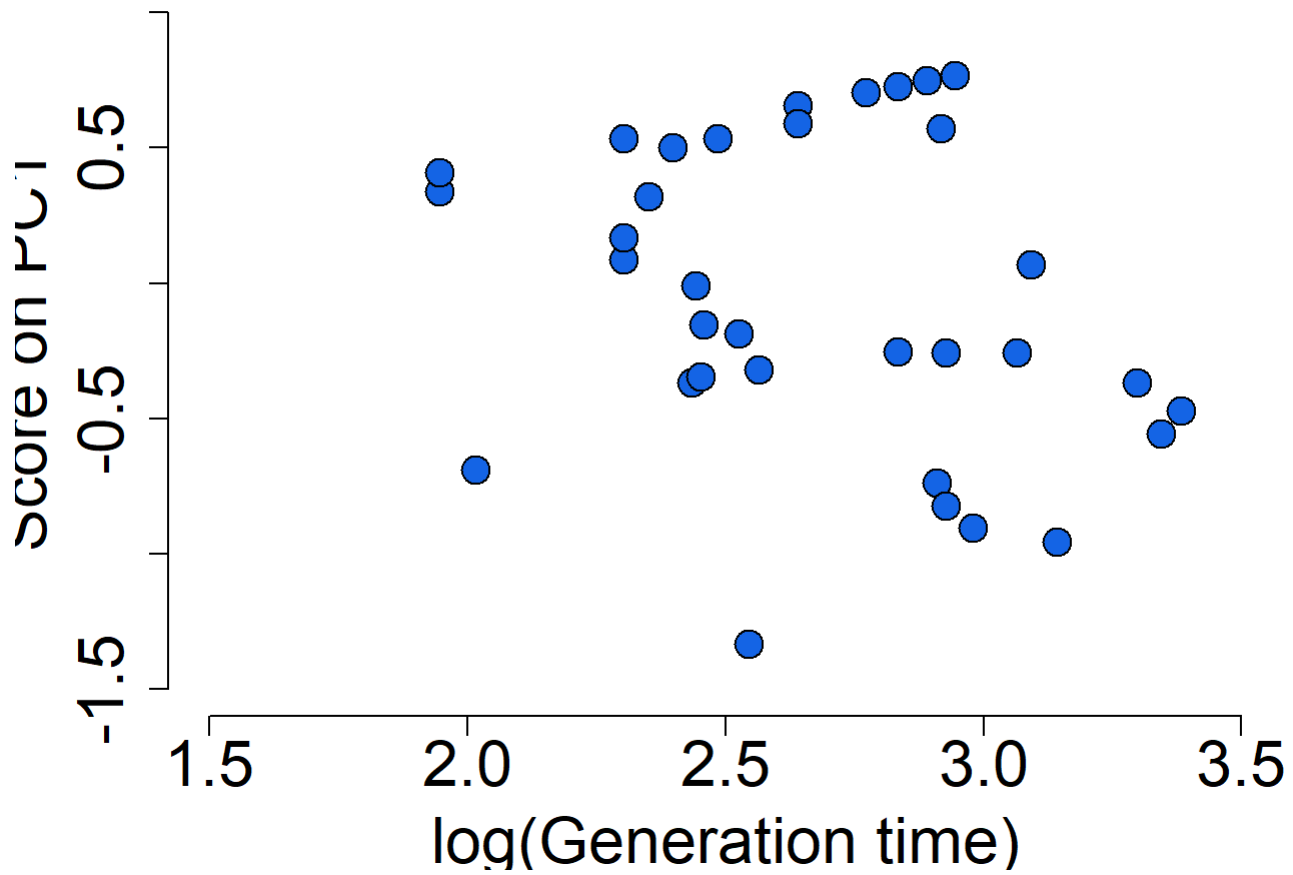
```
##      age_first_repro lifetime_adult breed_freq fec_one
## age_first_repro      1.0000000      -0.1336237  0.1148762 -0.1040642
## lifetime_adult      -0.1336237      1.0000000 -0.5993510  0.1193029
## breed_freq          0.1148762      -0.5993510  1.0000000  0.1107467
## fec_one             -0.1040642      0.1193029  0.1107467  1.0000000
```

Correlation with generation time

```
individual_score_petrel_reverse <- as.numeric(vegan::scores(test.pca_petrel1_reverse)$sites[,1])
summary(lm(individual_score_petrel_reverse ~ log(data_petrel$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_petrel_reverse ~ log(data_petrel$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.38214 -0.31964 -0.07465  0.41561  0.86866
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.9981     0.6977   1.431   0.162
## log(data_petrel$generation_time) -0.3732     0.2584  -1.444   0.158
##
## Residual standard error: 0.5726 on 34 degrees of freedom
## Multiple R-squared:  0.05779,    Adjusted R-squared:  0.03008
## F-statistic: 2.085 on 1 and 34 DF,  p-value: 0.1579
```

```
par(bty="n")
plot(log(data_petrel$generation_time), individual_score_petrel_reverse, pch=21, xlab="log(Genera
tion time)", ylab="Score on PC1", xlim=c(1.5,3.5), ylim=c(-1.5,1), bg=colors[14], cex=2, cex.axi
s=2, cex.lab=2)
```



Southern fulmar

```
data_fulmar<- data[which(data$species == "antarcticfulmar"),]  
data_fulmar$fec_one <- data_fulmar$fecundity + 1  
test.pca_fulmar1 <- rda(log(data_fulmar[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_fulmar1)
```

```
##
## Call:
## rda(X = log(data_fulmar[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained   4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1      PC2      PC3      PC4
## Eigenvalue      1.8995 1.1071 0.6854 0.30796
## Proportion Explained 0.4749 0.2768 0.1713 0.07699
## Cumulative Proportion 0.4749 0.7517 0.9230 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 5.136323
##
##
## Species scores
##
##           PC1      PC2      PC3      PC4
## age_first_repro -1.317  1.6008  1.5133 -0.09023
## lifetime_adult  2.002  1.2622 -0.3233 -0.94419
## breed_freq      -2.195 -0.8162 -0.2893 -1.01342
## fec_one         -1.403  1.5748 -1.4291  0.32326
##
##
## Site scores (weighted sums of species scores)
##
##           PC1      PC2      PC3      PC4
## 1290  0.320470  0.7020910  0.585354  0.235957
## 2100  0.402345 -0.2399767 -0.241616  0.075516
## 3100  0.871301  0.1857245  0.023748  0.762844
## 4100 -0.142810  0.5741452  0.632046  0.351223
## 5100  0.448438  0.0717192 -0.570428 -0.581875
## 6100 -0.033365  0.5788323  0.373442 -0.339798
## 7100 -0.119898 -0.1530982 -0.797518 -0.082867
## 8100 -0.077912 -0.6776497  0.503511 -0.122778
## 9100 -0.007050 -0.8254177  0.277873 -0.092835
## 10100 0.188288  0.5235654 -0.039512 -0.587406
## 11100 0.626975  0.0886987  0.146017  0.715524
## 1291 -0.355768  0.3517328  0.441440 -0.332275
## 1310 -0.418589  0.2902861 -0.001489 -0.020256
## 1410  0.500550  0.1289120 -0.072864 -0.396416
## 1510 -0.262230  0.2805506  0.123055 -0.386653
## 1610  0.262769  0.5513985  0.297933 -0.312088
## 1710  0.242837  0.4835777  0.372201  0.021182
```

## 1810	-0.007050	-0.8254177	0.277873	-0.092835
## 1910	0.424450	-0.0010427	-0.165900	0.791519
## 2010	-0.007050	-0.8254177	0.277873	-0.092835
## 2110	0.046442	0.4124093	0.119318	-0.306075
## 2210	0.293660	0.4131271	0.191235	-0.215028
## 2310	0.084306	-1.0159237	-0.013025	-0.054232
## 2410	0.400680	0.5385316	0.161961	-0.043331
## 2510	0.471527	0.3647706	0.539771	-0.304906
## 2610	-0.044454	-0.7474206	0.396973	-0.108640
## 2710	-0.122696	0.0005441	0.091261	-0.454295
## 2810	0.050950	0.1113244	0.154694	0.040183
## 2910	0.009483	0.2565670	0.147112	0.415575
## 3010	-0.622312	0.4082867	0.382115	0.216459
## 3110	0.151648	0.2615937	0.033963	-0.788507
## 3210	0.188075	0.2863596	0.108777	0.373846
## 3310	-0.058464	0.3711754	0.561586	-0.502195
## 3410	-0.034435	0.4383295	0.134854	-0.714727
## 3510	-0.416969	-0.0199165	-0.271740	0.303228
## 3610	0.841600	0.4629451	0.526166	0.967091
## 3710	0.204860	0.2651354	-0.080432	-0.508082
## 3810	0.695855	0.3069531	0.345501	0.209965
## 3910	-0.606818	0.3759773	0.332780	0.223006
## 4010	-0.416969	-0.0199165	-0.271740	0.303228
## 4110	-0.108179	-0.6145344	0.599887	-0.135568
## 4210	0.062533	0.1459420	-0.021809	0.437992
## 4310	-0.386702	-0.0830318	-0.368116	0.316018
## 4410	0.715745	0.1797864	0.332209	0.631257
## 4510	0.243614	0.3140574	-0.099982	-0.533717
## 4610	0.418286	0.0528058	-0.318238	-0.718006
## 4710	-0.135810	-0.5569145	0.687871	-0.147243
## 4810	-0.077912	-0.6776497	0.503511	-0.122778
## 4910	1.112822	0.1257723	-0.264918	0.963087
## 5010	-0.059879	0.3611573	0.078194	-0.353054
## 5110	0.336555	0.2214296	0.454419	0.171702
## 5210	-0.444600	0.0377034	-0.183756	0.291552
## 5310	-0.264569	-0.2884114	1.097868	-0.201651
## 5410	-0.172733	0.1906919	-0.151969	-0.429617
## 5510	1.055186	0.2136759	0.166070	0.961402
## 5610	0.114863	0.0205143	-0.526507	-0.354649
## 5710	0.051540	0.4048178	-0.024256	-0.442169
## 5810	-0.386702	-0.0830318	-0.368116	0.316018
## 5910	-0.637084	0.4390927	0.429155	0.210217
## 6010	0.377668	0.3522118	-0.203489	-0.358865
## 6110	0.309925	0.1800265	-0.101969	-0.722215
## 6210	0.386631	0.0178441	-0.258473	0.126107
## 6310	0.385952	0.1010319	0.075675	-0.254210
## 6410	0.035354	-0.9138436	0.142849	-0.074917
## 6510	-0.011306	-0.5097196	0.556516	-0.417433
## 6610	-0.059879	0.3611573	0.078194	-0.353054
## 6710	0.725618	0.0383059	0.026764	0.324571
## 6810	0.244825	0.1844616	-0.154237	-0.569855
## 6910	0.516743	0.1525040	0.145752	-0.074958

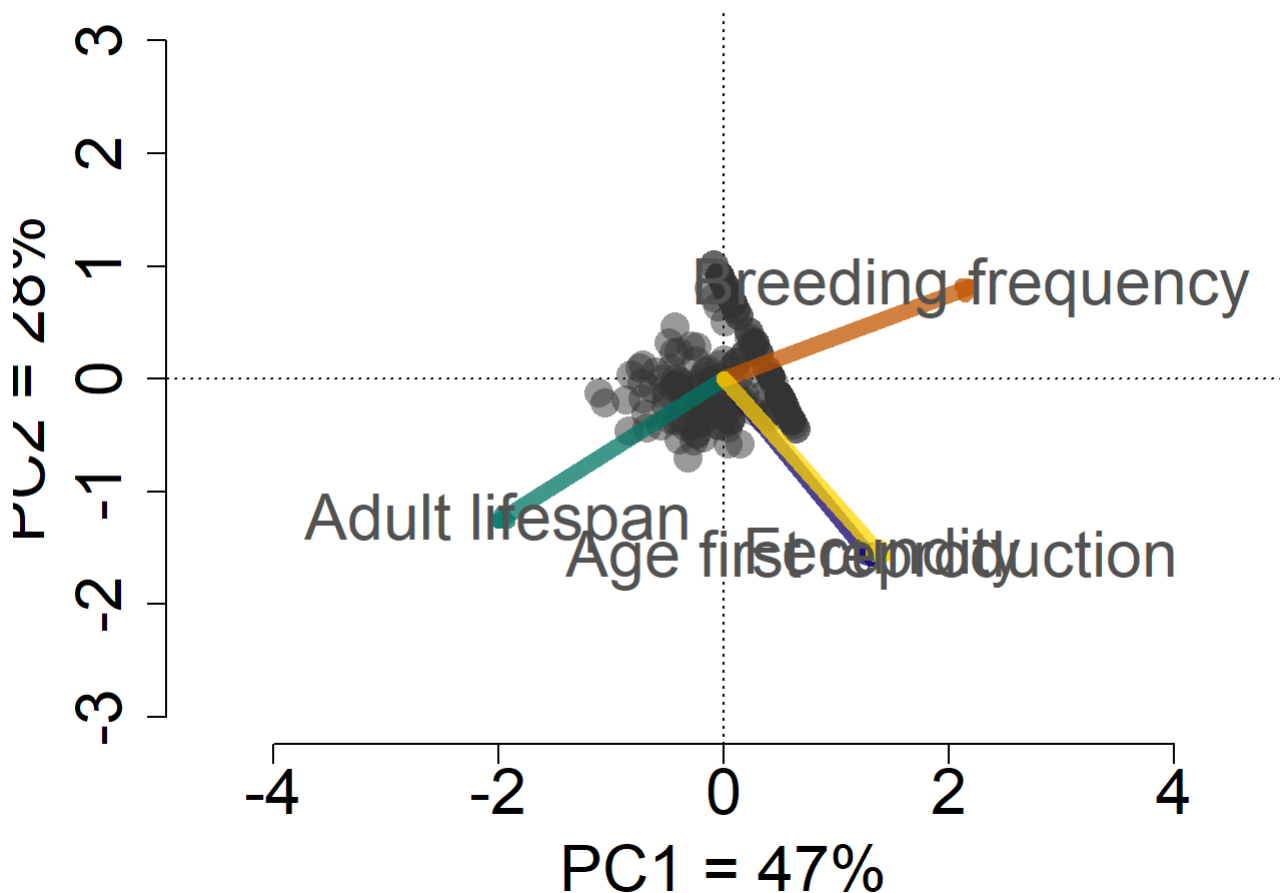
## 7010	0.487398	-0.3173582	-0.074486	0.264055
## 7110	0.047470	-0.1270537	-1.006408	-0.406732
## 7210	-0.606818	0.3759773	0.332780	0.223006
## 7310	0.132226	-0.8090288	0.099479	-0.356783
## 7410	-0.022841	0.2992793	-0.331577	-0.790468
## 7510	0.163661	-0.0649413	-0.343823	0.480724
## 7610	0.466854	-0.0894686	-0.300924	0.809437
## 7710	-0.493552	0.1397835	-0.027882	0.270867
## 7810	0.104555	0.0965392	-0.391130	-0.527966
## 7910	0.742515	-0.0971708	-0.068166	0.552481
## 8010	0.035354	-0.9138436	0.142849	-0.074917
## 8110	0.602489	-0.0592968	-0.234879	0.472489
## 8210	-0.439084	0.3330241	0.063771	-0.028916
## 8310	0.101361	0.2147914	-0.336250	-0.990115
## 8410	0.126258	0.0130558	-0.224723	0.464919
## 8510	-0.191369	0.1327826	-0.102583	-0.356710
## 8610	0.247431	-0.1600493	-0.241166	-0.707763
## 8810	0.444994	-0.2289323	0.060538	0.246137
## 8910	-0.136291	0.1879103	-0.480227	-0.412594
## 9010	-0.256372	-0.0479879	-0.518025	0.048290
## 9110	0.230227	0.0774000	-0.502376	-0.624708
## 9710	0.401335	0.0720671	-0.288857	0.363885
## 9810	0.512382	0.2260970	0.089163	0.345295
## 9910	-0.227166	-0.3664085	0.978769	-0.185847
## 10010	-0.135810	-0.5569145	0.687871	-0.147243
## 10110	0.214030	0.3917082	0.130245	-0.455503
## 10210	-0.637084	0.4390927	0.429155	0.210217
## 10310	0.265251	0.0554073	0.552258	-0.092083
## 10410	0.421550	0.3740558	0.193305	-0.060780
## 10510	-0.493552	0.1397835	-0.027882	0.270867
## 10610	-0.273437	-0.3192256	-0.728778	0.363879
## 10710	-0.119898	-0.1530982	-0.797518	-0.082867
## 10810	-0.001569	-0.1662372	-0.696991	-0.357291
## 10910	0.188440	0.3696893	-0.002566	-0.020430
## 11010	0.180145	-0.0294294	-0.668842	-0.110569
## 11110	0.108891	0.0651427	0.427713	0.274315
## 11210	-0.016095	0.2570553	-0.109970	-0.834343
## 11410	0.009483	0.2565670	0.147112	0.415575
## 11510	-0.176565	-0.2144109	-0.772148	0.082013
## 11610	0.290101	-0.2922709	0.620550	0.114614
## 11710	0.544868	-0.0022769	-0.383349	-0.164783
## 11810	0.235126	0.2020200	-0.176808	-0.495293
## 11910	-0.416969	-0.0199165	-0.271740	0.303228
## 12010	0.032074	-0.0153764	-0.136469	-0.711695
## 12110	0.402138	0.0270229	-0.539618	-0.182291
## 12210	-0.606818	0.3759773	0.332780	0.223006
## 12310	-0.493552	0.1397835	-0.027882	0.270867
## 12510	-0.347728	0.1425181	-0.227127	0.009687
## 12610	-0.079692	-0.1095962	-0.815519	-0.199852
## 12710	-0.535956	0.2282094	0.107142	0.252949
## 12810	-0.535956	0.2282094	0.107142	0.252949
## 1292	-0.535956	0.2282094	0.107142	0.252949

## 1301	-0.515462	0.1854713	0.041882	0.261610
## 1311	-0.044454	-0.7474206	0.396973	-0.108640
## 1321	-0.315841	-0.2307998	-0.593754	0.345960
## 1331	-0.493552	0.1397835	-0.027882	0.270867
## 1341	-0.573360	0.3062065	0.226241	0.237144
## 1351	-0.573360	0.3062065	0.226241	0.237144
## 1361	-0.590529	0.3420104	0.280913	0.229889
## 1371	0.721816	-0.1264618	-0.377427	0.308814
## 1381	-0.493552	0.1397835	-0.027882	0.270867
## 1391	0.431434	-0.4596522	0.280536	0.040362
## 1401	-0.007050	-0.8254177	0.277873	-0.092835
## 1411	-0.246418	-0.3262623	1.040071	-0.193982
## 1421	0.035354	-0.9138436	0.142849	-0.074917
## 1431	-0.044454	-0.7474206	0.396973	-0.108640
## 1441	-0.112967	0.1522405	-0.066599	-0.552477
## 1451	-0.416969	-0.0199165	-0.271740	0.303228
## 1461	0.823127	-0.0322365	0.215920	0.822015
## 1471	-0.535956	0.2282094	0.107142	0.252949
## 1481	0.084306	-1.0159237	-0.013025	-0.054232
## 1491	-0.416969	-0.0199165	-0.271740	0.303228
## 1501	-0.555208	0.2683556	0.168444	0.244814
## 1511	-0.224485	-0.4213058	-0.884652	0.384564
## 1521	-0.224485	-0.4213058	-0.884652	0.384564
## 1531	0.089822	-0.7206030	0.234503	-0.374701
## 1541	0.309286	0.0965287	0.014972	-0.233046
## 1551	0.111626	0.3219289	-0.229682	-0.642612
## 1561	-0.515462	0.1854713	0.041882	0.261610
## 1571	-0.515462	0.1854713	0.041882	0.261610
## 1581	-0.535956	0.2282094	0.107142	0.252949
## 1601	0.035354	-0.9138436	0.142849	-0.074917
## 1611	-0.535956	0.2282094	0.107142	0.252949
## 1621	0.035354	-0.9138436	0.142849	-0.074917
## 1641	-0.418589	0.2902861	-0.001489	-0.020256
## 1651	0.052419	-0.6426058	0.353602	-0.390506
## 1661	0.142306	-0.0791816	-0.136204	0.078786
## 1671	-0.273437	-0.3192256	-0.728778	0.363879
## 1681	-0.493552	0.1397835	-0.027882	0.270867
## 1691	0.243527	0.4881019	-0.119697	-0.688131
## 1701	0.234544	-0.2834299	-0.655551	-0.027621
## 1711	-0.176565	-0.2144109	-0.772148	0.082013
## 1721	0.336558	0.4261813	0.056821	-0.184324
## 1741	-0.315841	-0.2307998	-0.593754	0.345960
## 1751	-0.009488	0.3552426	-0.192926	-0.580055
## 1761	-0.315841	-0.2307998	-0.593754	0.345960
## 1771	-0.273437	-0.3192256	-0.728778	0.363879
## 1781	-0.077912	-0.6776497	0.503511	-0.122778
## 1791	0.680871	0.4343728	0.382025	0.388989
## 1811	0.561898	0.4198252	0.093179	-0.063811
## 1821	-0.416969	-0.0199165	-0.271740	0.303228
## 1831	-0.535956	0.2282094	0.107142	0.252949
## 1841	-0.416969	-0.0199165	-0.271740	0.303228
## 1851	-0.373146	0.1955232	-0.146189	-0.001054

```
## 1861 -0.108179 -0.6145344 0.599887 -0.135568
## 1871 -0.515462 0.1854713 0.041882 0.261610
```

```
# have the scores reversed to have the adult lifespan to the left
test.pca_fulmar1_reverse <- test.pca_fulmar1
test.pca_fulmar1_reverse$CA$u <- -(test.pca_fulmar1_reverse$CA$u)
test.pca_fulmar1_reverse$CA$v <- -(test.pca_fulmar1_reverse$CA$v)

# Figure
par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
plot(test.pca_fulmar1_reverse, scaling = scl, choices=c(1,2), type="n", ylim=c(-3,3), xlim=c(-3.5,3.5), main="", xlab="PC1 = 47%", ylab="PC2 = 28%", cex.lab=2, cex.axis=2)
with(data_fulmar, points(test.pca_fulmar1_reverse, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_fulmar1_reverse, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```

# Test for proportion of variance explained per axe
sum_all_fulmar <- summary(test.pca_fulmar1)

prop_fulmar <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_fulmar[j,i+1] <- as.numeric(abs(sum_all_fulmar$species[i,j])/sum(abs(sum_all_fulmar$species[i,])))
  }
}

prop_fulmar

```

```

##   PC      prim      adult      breed      freq
## 1  1 0.29130561 0.44171685 0.50882618 0.29663068
## 2  2 0.35405200 0.27855732 0.18919708 0.33291829
## 3  3 0.33468593 0.07135507 0.06706649 0.30211259
## 4  4 0.01995646 0.20837076 0.23491025 0.06833844

```

```

# Covariances
test <- as.matrix(scale(log(data_fulmar[,c(2,3,4,7)])))
cov(test)

```

```

##           age_first_repro lifetime_adult breed_freq   fec_one
## age_first_repro      1.0000000      -0.1546158   0.1877387   0.3301237
## lifetime_adult      -0.1546158       1.0000000  -0.6631023  -0.1006552
## breed_freq          0.1877387      -0.6631023   1.0000000   0.2851299
## fec_one             0.3301237      -0.1006552   0.2851299   1.0000000

```

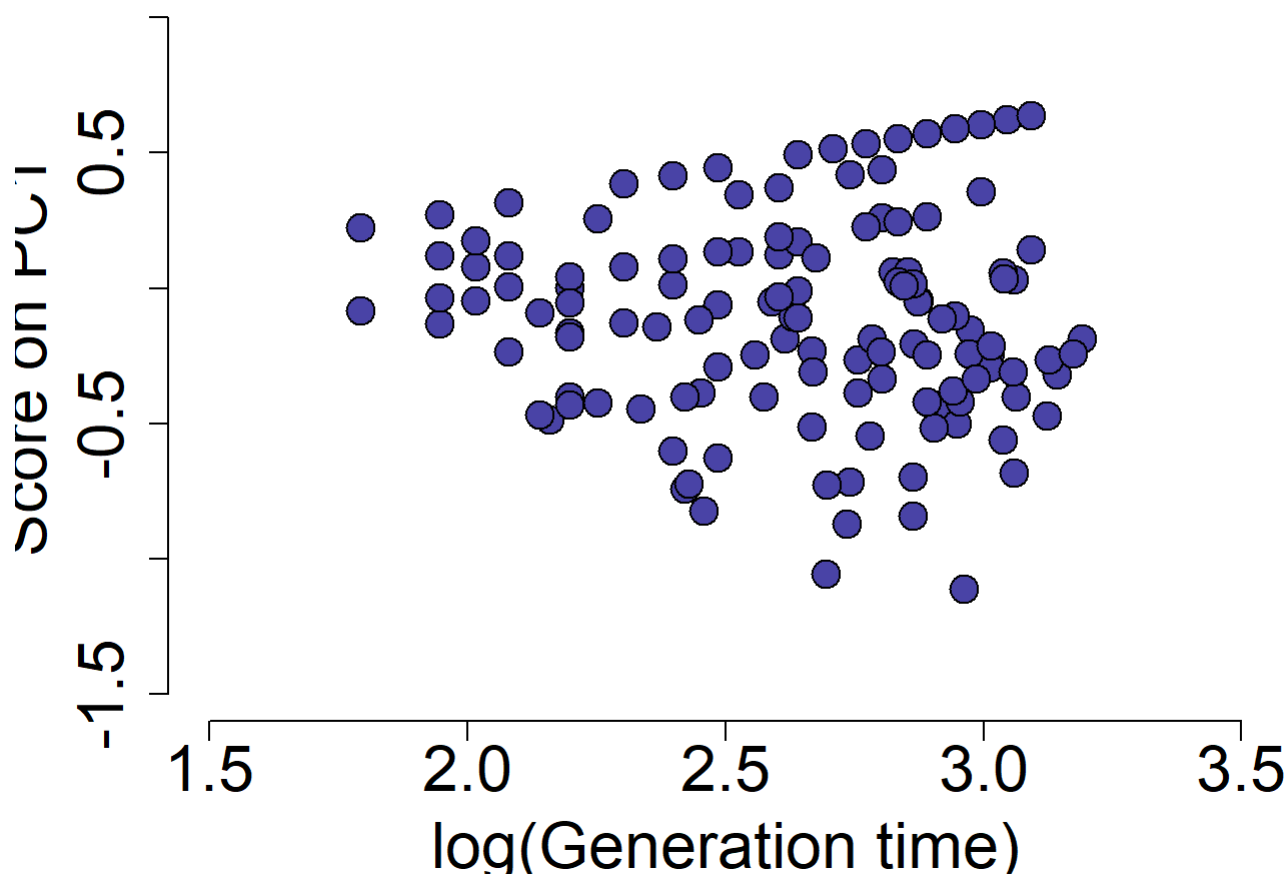
```

# Correlation with generation time
individual_score_fulmar_reverse <- as.numeric(vegan::scores(test.pca_fulmar1_reverse)$sites[,1])
summary(lm(individual_score_fulmar_reverse ~ log(data_fulmar$generation_time)))

```

```
##
## Call:
## lm(formula = individual_score_fulmar_reverse ~ log(data_fulmar$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.08871 -0.23838 -0.00536  0.28531  0.66920
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.15929    0.21312   0.747   0.456
## log(data_fulmar$generation_time) -0.06192    0.08205  -0.755   0.451
##
## Residual standard error: 0.3899 on 173 degrees of freedom
## Multiple R-squared:  0.003281,    Adjusted R-squared:  -0.00248
## F-statistic: 0.5696 on 1 and 173 DF,  p-value: 0.4515
```

```
par(bty="n")
plot(log(data_fulmar$generation_time), individual_score_fulmar_reverse, pch=21, xlab="log(Generation time)", ylab="Score on PC1", xlim=c(1.5,3.5), ylim=c(-1.5,1), bg=colors[1], cex=2, cex.axis=2, cex.lab=2)
```



Yellow-bellied marmot

```
data_marmots <- data[which(data$species == "yellowbelliedmarmots"),]  
data_marmots$fec_one <- data_marmots$fecundity + 1  
test.pca_marmots1 <- rda(log(data_marmots[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_marmots1)
```

```
##
## Call:
## rda(X = log(data_marmots[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained   4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      1.6139 1.0923 0.7786 0.5152
## Proportion Explained 0.4035 0.2731 0.1946 0.1288
## Cumulative Proportion 0.4035 0.6766 0.8712 1.0000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 5.885662
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro -1.603 -1.4414 -1.9980 0.1397
## lifetime_adult  -2.345  0.7538  0.6748  1.4617
## breed_freq       1.897 -1.7998  0.3292  1.3089
## fec_one          -1.519 -1.8906  1.4789 -0.7697
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 1340  -0.3496789 -0.313018 -0.248756 0.3942503
## 2137  -0.4116509 -0.254149 0.216295 0.4480475
## 3129   0.0905703 -0.017877 0.253862 0.0355201
## 4130   0.2350444 -0.138594 0.199711 -0.3262723
## 5128   0.2835472 -0.049378 0.101806 -0.2492679
## 6128  -0.3390970 -0.453950 -0.171555 0.1656955
## 7124  -0.7162943 0.087879 -0.624051 -0.0082626
## 8124   0.1666835 -0.264338 0.337700 -0.4348039
## 9123   0.4687718 0.291326 -0.272077 0.0447998
## 10123 0.0481991 -0.095816 0.339390 -0.0317497
## 11120 -0.3293703 -0.587137 -0.727207 0.1082254
## 12108 -0.2363826 -0.491083 -0.239974 -0.0170406
## 1348  -0.3496789 -0.313018 -0.248756 0.3942503
## 1430  -0.1034497 -0.246565 -0.508304 0.1940073
## 1538   0.1666835 -0.264338 0.337700 -0.4348039
## 1630  -0.1486032 0.003637 -0.175764 0.5198610
## 1737  -0.1647884 -0.704586 -0.518457 -0.3443537
```

##	1838	-0.0594391	0.007250	-0.257184	0.4160653
##	1929	-0.4486845	0.869021	-0.009306	-0.7672846
##	2029	-0.5198307	-0.173863	-0.183126	0.8157296
##	2138	0.1974227	-0.207796	0.275651	-0.3860015
##	2229	0.1849208	0.155672	0.063412	0.1853136
##	2329	-0.5198689	0.354710	-0.170230	-0.0402205
##	2428	0.4687718	0.291326	-0.272077	0.0447998
##	2528	0.3220931	0.096513	-0.650932	0.0848247
##	2626	0.0179336	0.234977	0.163007	0.5113634
##	2726	-0.0059847	-0.506956	0.011305	-0.4360408
##	2826	0.3017845	0.370632	-0.172482	0.3708496
##	2926	0.0574246	0.222210	-0.493077	0.6016014
##	3026	-0.1469824	-0.379845	0.058440	-0.0687292
##	3130	0.0905703	-0.017877	0.253862	0.0355201
##	3226	-0.0561083	-0.212691	-0.124993	0.0755450
##	3326	-0.5343503	0.516154	0.340896	-0.3826616
##	3426	0.2350444	-0.138594	0.199711	-0.3262723
##	3526	0.3017845	0.370632	-0.172482	0.3708496
##	3625	-0.1393123	-0.214659	0.578975	0.0163604
##	3725	0.3519081	0.076366	-0.036183	-0.1407362
##	3826	-0.6983605	0.675937	-0.264185	-0.1183804
##	3925	-0.0533261	-0.540831	-0.372006	-0.3175786
##	4026	-0.3679323	-0.087786	0.075234	0.6489267
##	4133	0.4687718	0.291326	-0.272077	0.0447998
##	4226	-0.2032853	-0.047521	0.533096	0.3504624
##	4326	0.3017845	0.370632	-0.172482	0.3708496
##	4426	0.4687718	0.291326	-0.272077	0.0447998
##	4526	0.3220931	0.096513	-0.650932	0.0848247
##	4625	-0.3120563	0.080124	0.551277	0.6790791
##	4725	0.0304354	-0.128490	0.375247	-0.0599517
##	4825	0.2835472	-0.049378	0.101806	-0.2492679
##	4925	0.3519081	0.076366	-0.036183	-0.1407362
##	5025	0.4687718	0.291326	-0.272077	0.0447998
##	5129	-0.5655482	0.654061	0.226588	-0.9528206
##	5225	-0.4952428	0.689742	0.199492	-0.1650771
##	5325	0.4687718	0.291326	-0.272077	0.0447998
##	5425	0.0680571	-0.059288	0.299306	-0.0002225
##	5525	-0.3075926	-0.164396	0.068713	0.4577570
##	5625	0.4687718	0.291326	-0.272077	0.0447998
##	5725	0.4687718	0.291326	-0.272077	0.0447998
##	5825	-0.3421215	-0.040848	0.652317	0.5308934
##	5925	0.4687718	0.291326	-0.272077	0.0447998
##	6025	-0.3179947	-0.415135	-0.214151	0.1991981
##	6129	0.4687718	0.291326	-0.272077	0.0447998
##	6224	0.2334236	0.244888	-0.034493	0.2623180
##	6324	0.4687718	0.291326	-0.272077	0.0447998
##	6424	-0.4540341	-0.380556	-0.114561	0.4188853
##	6523	-0.3317756	-0.625281	-0.418862	-0.0183039
##	6623	0.2052294	-0.118447	-0.415039	-0.1007114
##	6723	0.7195206	0.624362	0.375580	-0.0236232
##	6823	-0.1969223	-0.035817	0.520252	0.3605644
##	6923	0.0507440	-0.402609	-0.103204	-0.3459767

## 7020	0.3519081	0.076366	-0.036183	-0.1407362
## 7125	0.3519081	0.076366	-0.036183	-0.1407362
## 7220	-0.2832305	-0.243998	0.095987	0.3061229
## 7320	-0.6157373	0.113625	0.082920	-0.0074811
## 7420	-0.0296242	-0.012898	0.357565	0.1905044
## 7520	-0.4059247	0.489312	0.274701	-0.1322155
## 7620	0.0200049	-0.459151	-0.041156	-0.3947790
## 7720	-0.0925531	0.156161	0.309579	0.5262638
## 7820	-0.1826917	-0.392324	-0.348352	0.0682005
## 7920	-0.0284979	-0.548367	0.056749	-0.4717835
## 8020	0.4687718	0.291326	-0.272077	0.0447998
## 8125	-0.5710311	-0.268042	-0.079776	0.7344423
## 8219	-0.9064840	0.582868	-0.847940	-0.2862322
## 8320	0.0905703	-0.017877	0.253862	0.0355201
## 8420	-0.4276162	0.355786	0.409570	0.1700803
## 8519	0.1472991	0.086470	0.139353	0.1255843
## 8620	-0.0357573	0.136218	0.271384	0.4261223
## 8719	-0.7163062	0.696134	-0.706832	0.0467514
## 8820	0.3519081	0.076366	-0.036183	-0.1407362
## 8920	-0.4933453	0.338754	-0.164133	0.1868318
## 9020	-0.0059847	-0.506956	0.011305	-0.4360408
## 9124	0.0179336	0.234977	0.163007	0.5113634
## 9219	0.1472991	0.086470	0.139353	0.1255843
## 9319	0.1165599	0.029928	0.201401	0.0767819
## 9418	-0.4415511	0.423780	0.346615	-0.1887770
## 9518	0.1406939	-0.312143	0.390161	-0.4760657
## 9618	0.2835472	-0.049378	0.101806	-0.2492679
## 9719	-0.7404569	0.140820	-0.320423	-0.4336755
## 9819	-0.2591039	-0.199619	0.047286	0.3444270
## 9919	-0.0556138	-0.060703	0.410026	0.1492426
## 10019	-0.1826917	-0.392324	-0.348352	0.0682005
## 10124	0.1666835	-0.264338	0.337700	-0.4348039
## 10219	-0.3179947	-0.415135	-0.214151	0.1991981
## 10319	0.0481991	-0.095816	0.339390	-0.0317497
## 10419	-0.0672459	-0.082099	0.433506	0.1307752
## 10519	-0.1779236	0.175772	-0.255493	0.8191196
## 10619	-0.2821924	0.679352	0.091598	0.6301708
## 10719	0.1974227	-0.207796	0.275651	-0.3860015
## 10819	0.2350444	-0.138594	0.199711	-0.3262723
## 10918	0.2835472	-0.049378	0.101806	-0.2492679
## 11018	0.1974227	-0.207796	0.275651	-0.3860015
## 11121	-0.1162432	-0.323303	-0.003609	-0.0199269
## 11218	-0.0984796	-0.290629	-0.039465	0.0082752
## 11317	0.1974227	-0.207796	0.275651	-0.3860015
## 11418	-0.1041852	0.134764	0.333059	0.5077963
## 11518	-0.0379254	-0.254234	0.513236	-0.1684834
## 11618	-0.2602227	-0.050599	0.585563	0.4155632
## 11718	-0.0301187	-0.164885	-0.177454	0.1168068
## 11818	-0.6466229	0.027931	0.176712	0.0952750
## 11918	0.1666835	-0.264338	0.337700	-0.4348039
## 12018	-0.0071109	0.028513	0.312122	0.2262470
## 12118	-0.0137988	-0.209855	0.464535	-0.1301793

12218 -0.0840653 -0.597373 -0.309958 -0.3663809
12318 0.2180229 -0.041709 -0.919734 0.1132228
12417 0.1974227 -0.207796 0.275651 -0.3860015
12517 -0.5960730 0.214540 -0.016409 -0.1612042
12617 -0.0296242 -0.012898 0.357565 0.1905044
12717 -0.7745631 1.602633 -0.310536 -1.2468194
12817 -0.1365518 -0.049185 0.474842 0.2660981
12915 0.4687718 0.291326 -0.272077 0.0447998
13012 0.3220931 0.096513 -0.650932 0.0848247
13114 0.0507440 -0.402609 -0.103204 -0.3459767
13212 0.1368685 -0.244191 -0.277050 -0.2092430
13310 0.2835472 -0.049378 0.101806 -0.2492679
1349 -0.3130246 -0.459198 0.254688 0.0134661
1358 0.0883657 -0.333407 -0.179145 -0.2862474
1368 -0.0059847 -0.506956 0.011305 -0.4360408
1378 -0.0003037 -0.185032 0.437295 -0.1087541
1388 0.4687718 0.291326 -0.272077 0.0447998
1398 0.4687718 0.291326 -0.272077 0.0447998
1408 -0.2248056 -0.296927 0.076615 0.1535249
14114 0.1179787 0.258605 0.059623 0.4248428
14211 -0.3399593 -0.348345 0.210496 0.2160588
1438 -0.1663668 -0.029037 -0.139907 0.4916590
1448 -0.4059247 0.489312 0.274701 -0.1322155
1458 -0.2537793 -0.350222 0.135099 0.1075254
1468 -0.7408732 0.566952 -0.193005 -0.7129205
1478 0.0680571 -0.059288 0.299306 -0.0002225
1488 -0.1239746 -0.186447 0.548015 0.0407110
1498 0.4687718 0.291326 -0.272077 0.0447998
1508 0.3017845 0.370632 -0.172482 0.3708496
15114 -0.8111192 0.749754 -0.242937 -0.3471663
15211 -0.0658280 -0.177363 -0.584245 0.2537365
1539 0.0905703 -0.017877 0.253862 0.0355201
1548 -0.1572243 0.138857 0.377656 0.5790864
1557 0.0089217 0.132993 -0.395173 0.5245970
1568 0.3220931 0.096513 -0.650932 0.0848247
1578 -0.2248056 -0.296927 0.076615 0.1535249
1588 0.4687718 0.291326 -0.272077 0.0447998
1597 0.4687718 0.291326 -0.272077 0.0447998
1608 -0.2032853 -0.047521 0.533096 0.3504624
16114 0.7195206 0.624362 0.375580 -0.0236232
16211 -0.0854287 -0.040556 -0.204723 0.3748035
1637 0.0200049 -0.459151 -0.041156 -0.3947790
1648 0.2835472 -0.049378 0.101806 -0.2492679
1658 0.3519081 0.076366 -0.036183 -0.1407362
1668 0.3017845 0.370632 -0.172482 0.3708496
1678 -0.0533261 -0.540831 -0.372006 -0.3175786
1688 0.0304354 -0.128490 0.375247 -0.0599517
1698 -0.4498532 0.547165 0.246785 -0.4388238
1708 0.2350444 -0.138594 0.199711 -0.3262723
17114 0.0481991 -0.095816 0.339390 -0.0317497
17211 -0.4324541 -0.132017 0.159726 0.6603748
1738 0.2350444 -0.138594 0.199711 -0.3262723

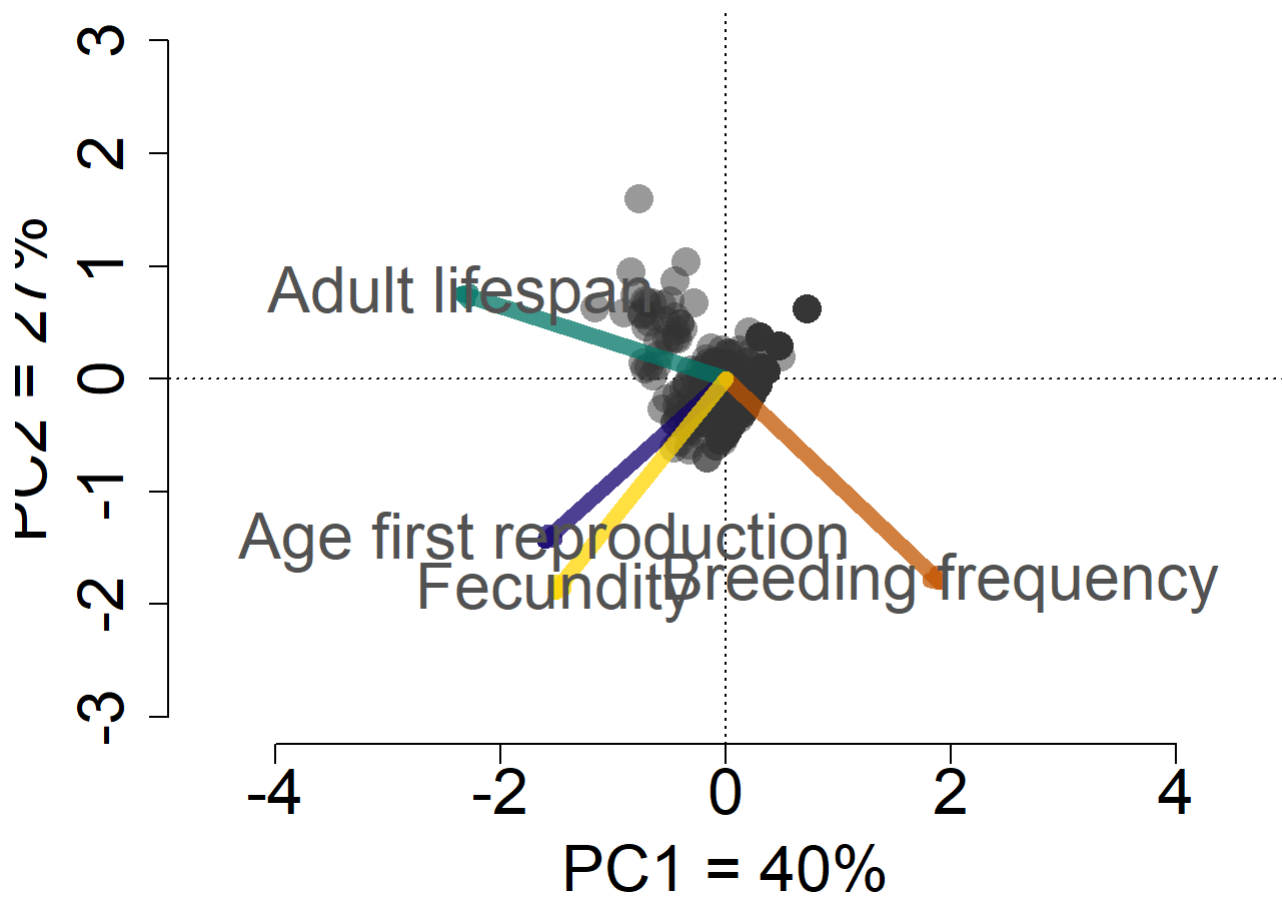
## 1748	-0.6840740	0.105876	-0.317647	-0.1072943
## 1758	-0.7122268	0.459248	-0.152268	-0.9127957
## 1768	0.1165599	0.029928	0.201401	0.0767819
## 1778	0.1368685	-0.244191	-0.277050	-0.2092430
## 1788	-0.6554981	0.563595	-0.266777	-0.8227316
## 1798	0.0883657	-0.333407	-0.179145	-0.2862474
## 1807	-0.4326902	-0.239643	-0.220109	0.6082677
## 18114	-0.1647884	-0.704586	-0.518457	-0.3443537
## 18211	-0.0262933	-0.232838	0.489756	-0.1500159
## 1839	0.1974227	-0.207796	0.275651	-0.3860015
## 1848	0.2350444	-0.138594	0.199711	-0.3262723
## 1858	0.3519081	0.076366	-0.036183	-0.1407362
## 1868	-0.7468633	0.586721	-0.166280	-0.1953848
## 1877	0.0382422	-0.039141	-0.315443	0.2253384
## 1886	0.4687718	0.291326	-0.272077	0.0447998
## 1896	-0.6074130	0.332337	-0.110106	-0.4160741
## 1906	0.2052294	-0.118447	-0.415039	-0.1007114
## 19113	-0.2106011	-0.110402	-0.050618	0.4214314
## 19210	0.4687718	0.291326	-0.272077	0.0447998
## 1936	-0.4540341	-0.380556	-0.114561	0.4188853
## 1946	-0.0533261	-0.540831	-0.372006	-0.3175786
## 1956	-0.0262933	-0.232838	0.489756	-0.1500159
## 1966	-0.2248056	-0.296927	0.076615	0.1535249
## 1976	0.1368685	-0.244191	-0.277050	-0.2092430
## 1986	-0.1285165	0.277609	0.266896	0.7561338
## 1996	0.1372999	-0.148922	-1.128234	0.1352500
## 2006	0.1472991	0.086470	0.139353	0.1255843
## 20113	0.4857932	0.194441	0.847368	-0.3946953
## 20210	-0.4646733	-0.609948	-0.593006	0.2392230
## 2036	-0.4394038	0.362986	0.401916	-0.0004250
## 2046	-0.1187882	-0.016510	0.438985	0.2943001
## 2056	0.0883657	-0.333407	-0.179145	-0.2862474
## 2066	0.2350444	-0.138594	0.199711	-0.3262723
## 2076	-0.8472788	0.951773	0.242660	-0.8849410
## 2086	0.1974227	-0.207796	0.275651	-0.3860015
## 2096	0.0204362	-0.363882	-0.892340	-0.0502860
## 21012	0.7195206	0.624362	0.375580	-0.0236232
## 21113	-0.0296242	-0.012898	0.357565	0.1905044
## 21211	-0.2350268	-0.315728	0.097247	0.1372974
## 2139	-0.3472617	-0.049764	0.033510	0.6817439
## 2146	-0.1740982	0.107819	0.411717	0.5522969
## 2156	0.1666835	-0.264338	0.337700	-0.4348039
## 2166	-0.6569134	0.677186	0.327085	-0.3254738
## 2176	-0.2337481	0.084045	0.479310	0.5890653
## 2186	-0.1663668	-0.029037	-0.139907	0.4916590
## 2196	-0.1239746	-0.186447	0.548015	0.0407110
## 2206	-0.0781270	-0.102114	0.455470	0.1135000
## 22113	-0.0357573	0.136218	0.271384	0.4261223
## 22210	0.1666835	-0.264338	0.337700	-0.4348039
## 2236	0.7195206	0.624362	0.375580	-0.0236232
## 2246	0.0883657	-0.333407	-0.179145	-0.2862474
## 2256	0.7195206	0.624362	0.375580	-0.0236232

##	2266	0.2350444	-0.138594	0.199711	-0.3262723
##	2276	0.1849208	0.155672	0.063412	0.1853136
##	2286	0.0680571	-0.059288	0.299306	-0.0002225
##	2296	-0.3543340	1.042570	-0.199756	-0.6174911
##	2306	0.3519081	0.076366	-0.036183	-0.1407362
##	23113	0.4687718	0.291326	-0.272077	0.0447998
##	23210	0.3220931	0.096513	-0.650932	0.0848247
##	2336	-0.0925531	0.156161	0.309579	0.5262638
##	2346	-0.0431193	-0.037721	0.384806	0.1690793
##	2356	0.2835472	-0.049378	0.101806	-0.2492679
##	2366	0.4687718	0.291326	-0.272077	0.0447998
##	2376	0.0089217	0.132993	-0.395173	0.5245970
##	2386	0.3220931	0.096513	-0.650932	0.0848247
##	2396	-0.1365518	-0.049185	0.474842	0.2660981
##	2406	0.7195206	0.624362	0.375580	-0.0236232
##	24112	-0.1824360	-0.058595	-0.107471	0.4661471
##	2429	0.0188787	0.076319	0.259661	0.2675089
##	2435	0.2041032	0.417023	-0.114222	0.5615765
##	2445	-0.0556138	-0.060703	0.410026	0.1492426
##	2455	-0.0840653	-0.597373	-0.309958	-0.3663809
##	2465	0.1181807	-0.353554	0.435604	-0.5118083
##	2475	0.0011150	0.043644	0.295517	0.2393068
##	2485	0.0507440	-0.402609	-0.103204	-0.3459767
##	2495	0.3188059	0.273747	0.946963	-0.0686454
##	2505	0.4687718	0.291326	-0.272077	0.0447998
##	25112	0.0507440	-0.402609	-0.103204	-0.3459767
##	2529	-0.0301187	-0.164885	-0.177454	0.1168068
##	2533	0.4687718	0.291326	-0.272077	0.0447998
##	2543	0.4687718	0.291326	-0.272077	0.0447998
##	2553	0.3220931	0.096513	-0.650932	0.0848247
##	2563	0.0507440	-0.402609	-0.103204	-0.3459767
##	2573	0.0200049	-0.459151	-0.041156	-0.3947790
##	2583	-0.0003037	-0.185032	0.437295	-0.1087541
##	2593	0.4687718	0.291326	-0.272077	0.0447998
##	2603	0.1666835	-0.264338	0.337700	-0.4348039
##	26110	0.1011593	-0.256669	-0.683841	-0.0723133
##	2627	0.2835472	-0.049378	0.101806	-0.2492679
##	2633	0.1974227	-0.207796	0.275651	-0.3860015
##	2643	0.0883657	-0.333407	-0.179145	-0.2862474
##	2653	-0.2784965	-0.395687	0.184992	0.0682838
##	2663	0.1011593	-0.256669	-0.683841	-0.0723133
##	2673	0.4687718	0.291326	-0.272077	0.0447998
##	2683	-0.0431193	-0.037721	0.384806	0.1690793
##	2693	0.3017845	0.370632	-0.172482	0.3708496
##	2703	-0.0494822	-0.049425	0.397650	0.1589772
##	27110	0.1406939	-0.312143	0.390161	-0.4760657
##	2727	0.0006205	-0.108343	-0.239502	0.1656092
##	2733	-1.1591586	0.633509	-0.705857	-0.6319156
##	2743	0.4687718	0.291326	-0.272077	0.0447998
##	2753	0.4687718	0.291326	-0.272077	0.0447998
##	2763	0.1165599	0.029928	0.201401	0.0767819
##	2773	0.4687718	0.291326	-0.272077	0.0447998

```
## 2783 -0.4270005 -0.330831 -0.169129 0.4618046
## 2793 0.0680571 -0.059288 0.299306 -0.0002225
## 2803 -0.0556138 -0.060703 0.410026 0.1492426
## 28110 0.0507440 -0.402609 -0.103204 -0.3459767
## 2827 -0.0970608 -0.061952 -0.181243 0.3563361
## 2833 0.0200049 -0.459151 -0.041156 -0.3947790
## 2842 -0.4198201 -0.561164 -0.380054 0.1877228
## 2853 -0.1278000 -0.118494 -0.119195 0.3075337
## 2863 -0.1826917 -0.392324 -0.348352 0.0682005
## 2873 0.1666835 -0.264338 0.337700 -0.4348039
## 2883 -0.0157044 -0.471629 -0.447947 -0.2578493
## 2893 -0.3791134 0.445002 0.311665 0.2470847
## 2903 -0.2226010 0.018602 0.509622 0.4752924
## 29110 -0.2248056 -0.296927 0.076615 0.1535249
## 2927 -0.0594391 0.007250 -0.257184 0.4160653
## 2933 -0.0157044 -0.471629 -0.447947 -0.2578493
## 2943 -0.6862372 0.507053 -0.204729 -0.8715339
## 2953 0.0200049 -0.459151 -0.041156 -0.3947790
## 2963 0.4687718 0.291326 -0.272077 0.0447998
## 2973 0.3017845 0.370632 -0.172482 0.3708496
## 2983 0.1974227 -0.207796 0.275651 -0.3860015
## 2993 0.1406939 -0.312143 0.390161 -0.4760657
## 3002 -0.0358243 0.260508 0.195070 0.6163280
## 30110 0.4687718 0.291326 -0.272077 0.0447998
```

Figure

```
par(mar=c(4,4,2,2), bty="n")
plot(test.pca_marmots1, scaling = scl, choices=c(1,2), type="n", main="", ylim=c(-3,3), xlim=c(
(-3.5,3.5), xlab="PC1 = 40%", ylab="PC2 = 27%", cex.axis=2, cex.lab=2)
with(data_marmots, points(test.pca_marmots1, display = "sites", col = col_points, scaling = scl,
pch = 16, cex=2))
g <- vegan::scores(test.pca_marmots1, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, c
ol_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```
# Test for proportion of variance explained per axe
sum_all_marmots <- summary(test.pca_marmots1)

prop_marmots <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_marmots[j,i+1] <- as.numeric(abs(sum_all_marmots$species[i,j])/sum(abs(sum_all_marmots
$species[i,])))
  }
}

prop_marmots
```

```
##      PC      prim      adult      breed      freq
## 1  1 0.30940066 0.4479398 0.35560660 0.2684198
## 2  2 0.27812375 0.1439812 0.33735223 0.3341632
## 3  3 0.38552483 0.1288843 0.06171224 0.2613823
## 4  4 0.02695076 0.2791946 0.24532893 0.1360348
```

Covariances

```
test <- as.matrix(scale(log(data_marmots[,c(2,3,4,7)])))
cov(test)
```

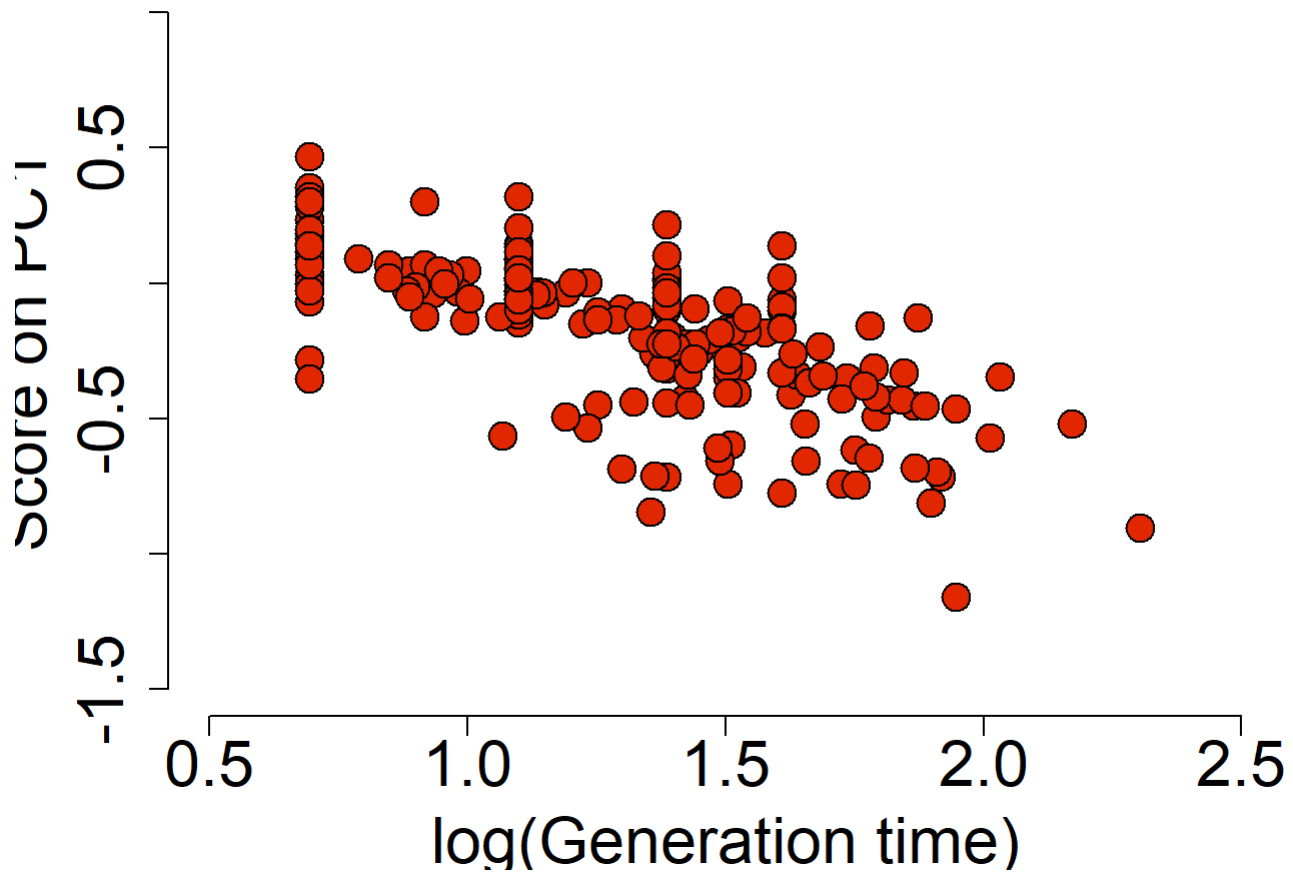
```
##           age_first_repro lifetime_adult      breed_freq      fec_one
## age_first_repro      1.0000000      0.1766588 -0.1065689812 0.2422620563
## lifetime_adult      0.1766588      1.0000000 -0.4238570847 0.2320096682
## breed_freq      -0.1065690      -0.4238571  1.0000000000 0.0001257311
## fec_one      0.2422621      0.2320097  0.0001257311 1.0000000000
```

Correlation with generation time

```
individual_score_marmots <- as.numeric(vegan::scores(test.pca_marmots1)$sites[,1])
summary(lm(individual_score_marmots ~ log(data_marmots$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_marmots ~ log(data_marmots$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.67899 -0.10199  0.00677  0.13614  0.47560
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      0.73519    0.03033   24.24  <2e-16 ***
## log(data_marmots$generation_time) -0.66700    0.02568  -25.97  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1886 on 299 degrees of freedom
## Multiple R-squared:  0.6928, Adjusted R-squared:  0.6918
## F-statistic: 674.4 on 1 and 299 DF,  p-value: < 2.2e-16
```

```
par(bty="n")
plot(log(data_marmots$generation_time), individual_score_marmots, pch=21, xlab="log(Generation time)", ylab="Score on PC1", xlim=c(0.5,2.5), ylim=c(-1.5,1), bg=colors[9], cex=2, cex.axis=2, cex.lab=2)
```



Great tits

```
data_gtits <- data[which(data$species == "greattits"),]  
data_gtits$fec_one <- data_gtits$fecundity + 1  
test.pca_gtits1 <- rda(log(data_gtits[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_gtits1)
```

```
##
## Call:
## rda(X = log(data_gtits[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total                2          1
## Unconstrained        2          1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1   PC2
## Eigenvalue      1.142 0.858
## Proportion Explained 0.571 0.429
## Cumulative Proportion 0.571 1.000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 6.05029
##
##
## Species scores
##
##           PC1   PC2
## age_first_repro 0.000 0.000
## lifetime_adult  3.233 2.802
## breed_freq      0.000 0.000
## fec_one         3.233 -2.802
##
##
## Site scores (weighted sums of species scores)
##
##           PC1   PC2
## 1350    0.005924 -0.202586
## 2140   -0.076810 -0.107133
## 3133   -0.442410  0.314676
## 4134    0.316261  0.040053
## 5130    0.005924 -0.202586
## 6130    0.046869 -0.249827
## 7126   -0.114012 -0.064212
## 8126    0.027422 -0.227390
## 9125   -0.159543 -0.011680
## 10125  -0.045356 -0.143423
## 11122   0.027422 -0.227390
## 12109  -0.076810 -0.107133
## 1359   -0.045356 -0.143423
## 1439   -0.076810 -0.107133
## 1540    0.064624 -0.270311
## 1638   -0.114012 -0.064212
## 1739  -0.218243  0.056045
```

## 1840	0.027422	-0.227390
## 1930	-0.076810	-0.107133
## 2030	0.027422	-0.227390
## 2147	-0.442410	0.314676
## 2230	-0.300976	0.151498
## 2330	-0.076810	-0.107133
## 2430	0.064624	-0.270311
## 2530	0.027422	-0.227390
## 2628	0.046869	-0.249827
## 2728	0.005924	-0.202586
## 2828	0.064624	-0.270311
## 2928	-0.045356	-0.143423
## 3027	-0.076810	-0.107133
## 3134	-0.018110	-0.174858
## 3227	0.046869	-0.249827
## 3327	-0.018110	-0.174858
## 3427	0.064624	-0.270311
## 3527	0.027422	-0.227390
## 3626	-0.442410	0.314676
## 3726	0.064624	-0.270311
## 3827	0.005924	-0.202586
## 3926	-0.018110	-0.174858
## 4027	-0.114012	-0.064212
## 4135	0.307191	0.050517
## 4227	-0.018110	-0.174858
## 4327	-0.045356	-0.143423
## 4427	0.005924	-0.202586
## 4527	0.477224	0.205724
## 4626	0.027422	-0.227390
## 4726	0.064624	-0.270311
## 4826	0.027422	-0.227390
## 4926	0.242212	0.125486
## 5026	-0.114012	-0.064212
## 5133	0.471476	0.212356
## 5226	-0.300976	0.151498
## 5326	0.080956	-0.289155
## 5426	0.005924	-0.202586
## 5526	0.135694	-0.352308
## 5626	0.096077	-0.306601
## 5726	-0.076810	-0.107133
## 5826	0.410823	0.282333
## 5926	0.266245	0.097758
## 6026	-0.300976	0.151498
## 6133	-0.442410	0.314676
## 6225	0.027422	-0.227390
## 6325	0.502534	0.425831
## 6425	-0.442410	0.314676
## 6524	0.110155	-0.322843
## 6624	0.214966	0.156922
## 6724	-0.045356	-0.143423
## 6824	0.080956	-0.289155
## 6924	-0.159543	-0.011680

## 7022	0.110155	-0.322843
## 7127	0.165758	0.213695
## 7222	0.046869	-0.249827
## 7322	0.080956	-0.289155
## 7422	0.027422	-0.227390
## 7522	0.135694	-0.352308
## 7622	-0.182088	0.615021
## 7722	0.254582	0.111214
## 7822	-0.018110	-0.174858
## 7922	0.046869	-0.249827
## 8022	0.046869	-0.249827
## 8127	-0.018110	-0.174858
## 8220	0.425944	0.264887
## 8322	0.110155	-0.322843
## 8422	0.027422	-0.227390
## 8520	0.080956	-0.289155
## 8622	0.027422	-0.227390
## 8720	-0.442410	0.314676
## 8822	0.027422	-0.227390
## 8922	0.123324	-0.338036
## 9022	0.385133	0.561281
## 9126	-0.076810	-0.107133
## 9220	-0.300976	0.151498
## 9320	-0.018110	-0.174858
## 9419	0.489365	0.441025
## 9519	0.242212	0.125486
## 9619	-0.045356	-0.143423
## 9720	-0.442410	0.314676
## 9820	0.324127	0.382358
## 9920	0.005924	-0.202586
## 10020	-0.045356	-0.143423
## 10126	0.005924	-0.202586
## 10220	0.100779	0.288664
## 10320	0.460166	0.474713
## 10420	0.124812	0.260936
## 10520	-0.114012	-0.064212
## 10620	0.229043	0.140680
## 10720	0.446712	0.240926
## 10820	0.269389	0.445511
## 10919	0.005924	-0.202586
## 11019	0.005924	-0.202586
## 11123	0.110155	-0.322843
## 11219	0.005924	-0.202586
## 11318	-0.076810	-0.107133
## 11419	-0.114012	-0.064212
## 11519	0.042079	0.356389
## 11619	0.514904	0.411559
## 11719	0.287743	0.072955
## 11819	-0.114012	-0.064212
## 11919	-0.442410	0.314676
## 12019	-0.018110	-0.174858
## 12119	-0.159543	-0.011680

12219 -0.045356 -0.143423
12319 -0.114012 -0.064212
12418 -0.076810 -0.107133
12518 0.520819 0.404735
12618 0.027422 -0.227390
12718 0.005924 -0.202586
12818 0.183512 0.193211
12916 0.078234 0.915365
13013 -0.442410 0.314676
13115 0.402820 0.291567
13213 -0.076810 -0.107133
13311 0.110155 -0.322843
13410 0.027422 -0.227390
13510 0.418524 0.273449
1369 0.346822 0.356173
1379 -0.218243 0.056045
1389 -0.159543 -0.011680
1399 0.520819 0.404735
1409 -0.045356 -0.143423
14115 0.027422 -0.227390
14212 0.183512 0.193211
14310 -0.159543 -0.011680
1449 0.096077 -0.306601
1459 0.277277 0.085030
1469 -0.018110 -0.174858
1479 -0.182088 0.615021
1489 0.110155 -0.322843
1499 -0.442410 0.314676
1509 -0.159543 -0.011680
15115 -0.442410 0.314676
15212 0.124812 0.260936
15310 -0.442410 0.314676
1549 -0.159543 -0.011680
1558 -0.442410 0.314676
1569 -0.076810 -0.107133
1579 -0.442410 0.314676
1589 0.410823 0.282333
1598 0.376736 0.321661
1609 -0.300976 0.151498
16115 -0.159543 -0.011680
16212 -0.300976 0.151498
1639 0.183512 0.193211
1649 0.311757 0.396630
1659 -0.442410 0.314676
1669 -0.182088 0.615021
1679 -0.442410 0.314676
1689 -0.076810 -0.107133
1699 0.005924 -0.202586
1709 -0.300976 0.151498
17115 -0.114012 -0.064212
17212 -0.018110 -0.174858
17310 0.357289 0.344098

## 1749	-0.076810	-0.107133
## 1759	-0.442410	0.314676
## 1769	-0.442410	0.314676
## 1779	-0.029810	0.790711
## 1789	-0.442410	0.314676
## 1799	-0.076810	-0.107133
## 1808	0.277277	0.085030
## 18115	0.287743	0.072955
## 18212	-0.045356	-0.143423
## 18310	0.242212	0.125486
## 1849	0.027422	-0.227390
## 1859	0.073532	0.320100
## 1869	0.287743	0.072955
## 1878	0.266245	0.097758
## 1887	0.165758	0.213695
## 1897	-0.300976	0.151498
## 1907	-0.045356	-0.143423
## 19114	0.046869	-0.249827
## 19211	-0.442410	0.314676
## 1937	-0.076810	-0.107133
## 1947	-0.442410	0.314676
## 1957	-0.018110	-0.174858
## 1967	-0.045356	-0.143423
## 1977	-0.045356	-0.143423
## 1987	-0.076810	-0.107133
## 1997	-0.076810	-0.107133
## 2007	0.005924	-0.202586
## 20114	0.027422	-0.227390
## 20211	-0.114012	-0.064212
## 2037	0.287743	0.072955
## 2047	-0.045356	-0.143423
## 2057	-0.018110	-0.174858
## 2067	0.004877	0.399311
## 2077	-0.218243	0.056045
## 2087	-0.018110	-0.174858
## 2097	-0.076810	-0.107133
## 21013	0.005924	-0.202586
## 21114	-0.218243	0.056045
## 21212	0.005924	-0.202586
## 21310	0.005924	-0.202586
## 2148	-0.300976	0.151498
## 2157	0.027422	-0.227390
## 2167	0.005924	-0.202586
## 2177	0.005924	-0.202586
## 2187	0.253057	0.464355
## 2197	0.254582	0.111214
## 2207	0.005924	-0.202586
## 22114	0.005924	-0.202586
## 22211	-0.442410	0.314676
## 2237	0.046869	-0.249827
## 2247	0.046869	-0.249827
## 2257	0.027422	-0.227390

##	2267	0.046869	-0.249827
##	2277	-0.018110	-0.174858
##	2287	0.214966	0.156922
##	2297	0.005924	-0.202586
##	2307	0.005924	-0.202586
##	23114	0.346822	0.356173
##	23211	0.254582	0.111214
##	2337	0.266245	0.097758
##	2347	0.254582	0.111214
##	2357	-0.076810	-0.107133
##	2367	0.005924	-0.202586
##	2377	0.027422	-0.227390
##	2387	-0.045356	-0.143423
##	2397	-0.159543	-0.011680
##	2407	-0.045356	-0.143423
##	24113	-0.018110	-0.174858
##	24210	-0.018110	-0.174858
##	2436	-0.076810	-0.107133
##	2446	-0.442410	0.314676
##	2456	-0.018110	-0.174858
##	2466	-0.442410	0.314676
##	2476	0.005924	-0.202586
##	2486	-0.159543	-0.011680
##	2496	0.005924	-0.202586
##	2506	0.005924	-0.202586
##	25113	-0.018110	-0.174858
##	25210	0.064624	-0.270311
##	2534	-0.018110	-0.174858
##	2544	0.005924	-0.202586
##	2554	0.046869	-0.249827
##	2564	-0.114012	-0.064212
##	2574	-0.442410	0.314676
##	2584	0.064624	-0.270311
##	2594	0.214966	0.156922
##	2604	0.046869	-0.249827
##	26111	0.046869	-0.249827
##	2629	-0.159543	-0.011680
##	2634	-0.076810	-0.107133
##	2644	-0.442410	0.314676
##	2654	-0.114012	-0.064212
##	2664	0.027422	-0.227390
##	2674	-0.018110	-0.174858
##	2684	0.027422	-0.227390
##	2694	-0.045356	-0.143423
##	2704	-0.018110	-0.174858
##	27111	0.297699	0.061469
##	2729	0.046869	-0.249827
##	2734	-0.018110	-0.174858
##	2744	-0.300976	0.151498
##	2754	-0.018110	-0.174858
##	2764	0.027422	-0.227390
##	2774	-0.442410	0.314676

2784 -0.159543 -0.011680
2794 -0.159543 -0.011680
2804 0.005924 -0.202586
28111 0.027422 -0.227390
2829 0.046869 -0.249827
2834 0.619245 0.484555
2843 0.005924 -0.202586
2854 -0.045356 -0.143423
2864 0.080956 -0.289155
2874 0.005924 -0.202586
2884 -0.018110 -0.174858
2894 -0.018110 -0.174858
2904 -0.159543 -0.011680
29111 0.005924 -0.202586
2929 0.027422 -0.227390
2934 0.027422 -0.227390
2944 -0.076810 -0.107133
2954 -0.442410 0.314676
2964 0.005924 -0.202586
2974 0.005924 -0.202586
2984 0.027422 -0.227390
2994 -0.300976 0.151498
3003 -0.045356 -0.143423
30111 -0.018110 -0.174858
3028 0.005924 -0.202586
3033 0.316261 0.040053
3043 0.046869 -0.249827
3053 -0.018110 -0.174858
3063 -0.218243 0.056045
3073 0.027422 -0.227390
3082 0.027422 -0.227390
3093 -0.045356 -0.143423
31010 -0.300976 0.151498
31110 0.046869 -0.249827
31210 -0.076810 -0.107133
3135 -0.182088 0.615021
3143 -0.114012 -0.064212
3153 -0.076810 -0.107133
3163 0.601599 0.311535
3173 0.229043 0.140680
3183 0.046869 -0.249827
3193 0.440022 0.248645
3203 0.165758 0.213695
32110 -0.018110 -0.174858
3228 -0.442410 0.314676
3233 0.266245 0.097758
3243 -0.218243 0.056045
3253 -0.114012 -0.064212
3263 0.005924 -0.202586
3273 0.333275 0.020423
3283 0.064624 -0.270311
3293 0.214966 0.156922

## 3303	0.465561	0.219180
## 33110	0.027422	-0.227390
## 3328	0.005924	-0.202586
## 3333	-0.045356	-0.143423
## 3343	-0.114012	-0.064212
## 3352	-0.018110	-0.174858
## 3363	0.418524	0.273449
## 3373	0.027422	-0.227390
## 3383	-0.442410	0.314676
## 3393	-0.114012	-0.064212
## 3403	0.027422	-0.227390
## 34110	0.254582	0.111214
## 3428	-0.018110	-0.174858
## 3433	-0.442410	0.314676
## 3443	-0.076810	-0.107133
## 3453	-0.442410	0.314676
## 3463	0.027422	-0.227390
## 3473	0.046869	-0.249827
## 3483	0.046869	-0.249827
## 3493	-0.076810	-0.107133
## 3503	-0.018110	-0.174858
## 35110	0.183512	0.193211
## 3528	-0.045356	-0.143423
## 3533	0.046869	-0.249827
## 3543	-0.045356	-0.143423
## 3553	0.005924	-0.202586
## 3563	-0.218243	0.056045
## 3573	0.005924	-0.202586
## 3583	0.254582	0.111214
## 3593	-0.045356	-0.143423
## 3603	-0.218243	0.056045
## 36110	-0.018110	-0.174858
## 3627	-0.114012	-0.064212
## 3633	0.080956	-0.289155
## 3643	-0.018110	-0.174858
## 3653	-0.045356	-0.143423
## 3663	-0.045356	-0.143423
## 3673	-0.076810	-0.107133
## 3683	-0.442410	0.314676
## 3693	-0.018110	-0.174858
## 3703	-0.018110	-0.174858
## 37110	-0.045356	-0.143423
## 3727	-0.442410	0.314676
## 3733	-0.442410	0.314676
## 3743	-0.442410	0.314676
## 3753	0.080956	-0.289155
## 3763	0.005924	-0.202586
## 3773	0.335790	0.368902
## 3783	0.298588	0.411823
## 3793	0.214966	0.156922
## 3803	0.376736	0.321661
## 38110	0.199844	0.174368

3828 0.027422 -0.227390
3833 -0.442410 0.314676
3843 0.453191 0.233452
3853 0.357289 0.344098
3863 -0.218243 0.056045
3873 -0.114012 -0.064212
3883 0.080956 -0.289155
3893 -0.076810 -0.107133
3903 -0.159543 -0.011680
39110 -0.159543 -0.011680
3927 -0.442410 0.314676
3933 -0.114012 -0.064212
3943 -0.076810 -0.107133
3953 -0.159543 -0.011680
3963 -0.045356 -0.143423
3973 0.027422 -0.227390
3983 -0.045356 -0.143423
3993 -0.159543 -0.011680
4003 -0.040655 0.451842
40110 -0.045356 -0.143423
4028 0.165758 0.213695
4033 -0.442410 0.314676
4043 -0.159543 -0.011680
4053 -0.300976 0.151498
4063 0.324127 0.382358
4073 -0.076810 -0.107133
4083 0.146310 0.236133
4093 0.004877 0.399311
41010 -0.442410 0.314676
41110 -0.442410 0.314676
41210 -0.442410 0.314676
4136 -0.159543 -0.011680
4142 -0.218243 0.056045
4153 -0.182088 0.615021
4162 0.046869 -0.249827
4173 -0.159543 -0.011680
4183 -0.076810 -0.107133
4193 0.143077 0.591243
4203 -0.442410 0.314676
42110 -0.018110 -0.174858
4228 -0.442410 0.314676
4233 -0.442410 0.314676
4243 -0.114012 -0.064212
4253 -0.442410 0.314676
4263 -0.442410 0.314676
4273 -0.442410 0.314676
4283 0.027422 -0.227390
4293 -0.442410 0.314676
4303 -0.442410 0.314676
43110 -0.159543 -0.011680
4328 0.046869 -0.249827
4333 -0.114012 -0.064212

4343 -0.045356 -0.143423
4353 -0.218243 0.056045
4363 0.005924 -0.202586
4373 -0.442410 0.314676
4383 -0.300976 0.151498
4393 0.027422 -0.227390
4403 -0.045356 -0.143423
44110 0.199844 0.174368
4428 0.254582 0.111214
4433 0.005924 -0.202586
4443 -0.018110 -0.174858
4453 -0.114012 -0.064212
4463 -0.018110 -0.174858
4473 -0.045356 -0.143423
4483 -0.442410 0.314676
4493 -0.159543 -0.011680
4503 -0.076810 -0.107133
45110 -0.114012 -0.064212
4528 0.183512 0.193211
4533 0.348978 0.002305
4543 0.341278 0.011190
4553 -0.159543 -0.011680
4563 -0.300976 0.151498
4573 -0.442410 0.314676
4583 -0.114012 -0.064212
4593 -0.182088 0.615021
4603 -0.442410 0.314676
46110 -0.018110 -0.174858
4627 -0.045356 -0.143423
4633 0.027422 -0.227390
4643 0.005924 -0.202586
4653 -0.045356 -0.143423
4663 0.064624 -0.270311
4673 0.096077 -0.306601
4683 -0.076810 -0.107133
4693 -0.018110 -0.174858
4703 -0.045356 -0.143423
47110 0.453191 0.233452
4727 -0.018110 -0.174858
4733 0.297699 0.061469
4743 -0.114012 -0.064212
4752 0.254582 0.111214
4763 -0.018110 -0.174858
4773 -0.045356 -0.143423
4783 -0.018110 -0.174858
4793 -0.159543 -0.011680
4803 -0.018110 -0.174858
48110 -0.018110 -0.174858
4827 0.277277 0.085030
4833 0.135694 -0.352308
4843 0.168855 -0.390568
4853 0.527240 0.148018

4863 -0.442410 0.314676
4873 0.027422 -0.227390
4883 0.046869 -0.249827
4893 0.471476 0.212356
4903 0.396016 -0.051964
49110 0.027422 -0.227390
4927 0.046869 -0.249827
4933 0.394490 0.301177
4943 0.027422 -0.227390
4953 0.513479 0.163895
4963 -0.018110 -0.174858
4973 -0.076810 -0.107133
4983 0.005924 -0.202586
4993 -0.018110 -0.174858
5003 0.005924 -0.202586
50110 0.064624 -0.270311
5027 0.080956 -0.289155
5033 0.046869 -0.249827
5043 -0.018110 -0.174858
5053 0.064624 -0.270311
5063 0.064624 -0.270311
5073 0.229043 0.140680
5083 0.046869 -0.249827
5093 0.027422 -0.227390
51010 0.046869 -0.249827
51110 0.199844 0.174368
51210 -0.018110 -0.174858
5134 0.005924 -0.202586
5143 -0.300976 0.151498
5153 -0.218243 0.056045
5163 0.418524 0.273449
5173 0.493556 0.186880
5183 0.266245 0.097758
5193 -0.442410 0.314676
5203 -0.442410 0.314676
52110 -0.018110 -0.174858
5227 0.383645 -0.037692
5233 -0.076810 -0.107133
5243 0.242212 0.125486
5253 -0.114012 -0.064212
5263 -0.114012 -0.064212
5273 0.333275 0.020423
5283 -0.159543 -0.011680
5293 0.080956 -0.289155
5303 0.277277 0.085030
53110 -0.045356 -0.143423
5327 0.277277 0.085030
5333 -0.018110 -0.174858
5343 0.073532 0.320100
5353 0.046869 -0.249827
5363 -0.018110 -0.174858
5372 0.316261 0.040053

## 5383	0.297699	0.061469
## 5393	0.027422	-0.227390
## 5403	0.254582	0.111214
## 54110	-0.018110	-0.174858
## 5427	-0.018110	-0.174858
## 5433	0.452163	0.483946
## 5443	0.453191	0.233452
## 5452	-0.076810	-0.107133
## 5463	0.027422	-0.227390
## 5473	0.005924	-0.202586
## 5483	0.073532	0.320100
## 5493	-0.018110	-0.174858
## 5503	-0.076810	-0.107133
## 55110	0.146310	0.236133
## 5527	-0.045356	-0.143423
## 5533	-0.045356	-0.143423
## 5543	-0.018110	-0.174858
## 5553	0.307191	0.050517
## 5563	-0.442410	0.314676
## 5573	0.425944	0.264887
## 5583	0.535924	0.137999
## 5593	0.502534	0.425831
## 5603	0.426079	0.514040
## 56110	0.277277	0.085030
## 5627	0.046869	-0.249827
## 5633	0.064624	-0.270311
## 5643	-0.218243	0.056045
## 5653	-0.018110	-0.174858
## 5663	-0.076810	-0.107133
## 5673	-0.076810	-0.107133
## 5683	-0.018110	-0.174858
## 5693	0.410823	0.282333
## 5703	-0.045356	-0.143423
## 57110	-0.442410	0.314676
## 5727	-0.018110	-0.174858
## 5733	0.693426	0.556971
## 5743	0.311757	0.396630
## 5753	-0.018110	-0.174858
## 5763	-0.442410	0.314676
## 5773	-0.218243	0.056045
## 5783	-0.442410	0.314676
## 5793	0.376736	0.321661
## 5803	0.508813	0.418587
## 58110	-0.218243	0.056045
## 5827	0.376736	0.321661
## 5833	-0.045356	-0.143423
## 5843	0.287743	0.072955
## 5853	-0.114012	-0.064212
## 5863	0.183512	0.193211
## 5873	-0.018110	-0.174858
## 5883	-0.159543	-0.011680
## 5893	-0.018110	-0.174858

## 5903	0.266245	0.097758
## 59110	-0.099355	0.519567
## 5927	0.367244	0.332612
## 5933	0.367244	0.332612
## 5943	0.124812	0.260936
## 5953	0.027422	-0.227390
## 5963	0.183512	0.193211
## 5973	0.064624	-0.270311
## 5983	0.367244	0.332612
## 5993	-0.114012	-0.064212
## 6003	0.100779	0.288664
## 60110	-0.114012	-0.064212
## 6027	-0.300976	0.151498
## 6033	0.215855	0.507276
## 6043	0.446712	0.240926
## 6053	-0.442410	0.314676
## 6063	-0.114012	-0.064212
## 6073	0.548065	0.373300
## 6083	-0.076810	-0.107133
## 6093	0.453191	0.233452
## 61010	0.537599	0.385375
## 61110	-0.159543	-0.011680
## 61210	0.425944	0.264887
## 6134	0.316261	0.040053
## 6143	-0.218243	0.056045
## 6153	0.005924	-0.202586
## 6163	-0.114012	-0.064212
## 6173	-0.045356	-0.143423
## 6183	0.394490	0.301177
## 6193	-0.442410	0.314676
## 6203	0.459469	0.226208
## 62110	-0.018110	-0.174858
## 6226	0.440022	0.248645
## 6233	-0.076810	-0.107133
## 6243	0.346822	0.356173
## 6253	-0.159543	-0.011680
## 6263	-0.300976	0.151498
## 6273	-0.076810	-0.107133
## 6283	0.064624	-0.270311
## 6293	-0.040655	0.451842
## 6303	0.376736	0.321661
## 63110	0.433104	0.256626
## 6326	0.440022	0.248645
## 6333	-0.442410	0.314676
## 6343	-0.045356	-0.143423
## 6353	-0.114012	-0.064212
## 6363	0.046869	-0.249827
## 6373	0.214966	0.156922
## 6383	0.046869	-0.249827
## 6393	-0.442410	0.314676
## 6403	-0.018110	-0.174858
## 64110	0.046869	-0.249827

```
## 6426 -0.114012 -0.064212
## 6433 -0.442410 0.314676
## 6443 0.046869 -0.249827
## 6453 -0.045356 -0.143423
## 6463 0.005924 -0.202586
## 6473 -0.018110 -0.174858
## 6483 -0.018110 -0.174858
## 6493 -0.218243 0.056045
## 6503 -0.218243 0.056045
## 65110 -0.114012 -0.064212
## 6525 0.027422 -0.227390
## 6533 -0.300976 0.151498
## 6542 -0.159543 -0.011680
## 6553 -0.018110 -0.174858
## 6563 -0.218243 0.056045
## 6573 0.459469 0.226208
## 6583 -0.045356 -0.143423
## 6593 0.123324 -0.338036
## 6603 -0.045356 -0.143423
## 66110 0.183512 0.193211
## 6625 -0.045356 -0.143423
## 6633 0.333275 0.020423
## 6643 0.005924 -0.202586
## 6653 0.254582 0.111214
## 6663 -0.442410 0.314676
## 6673 -0.018110 -0.174858
## 6683 0.183512 0.193211
## 6693 0.229043 0.140680
## 6703 0.046869 -0.249827
## 67110 -0.442410 0.314676
```

```
# have the scores reversed to have the adult lifespan to the left
```

```
test.pca_gtits1_reverse <- test.pca_gtits1
```

```
test.pca_gtits1_reverse$CA$u <- -(test.pca_gtits1_reverse$CA$u)
```

```
test.pca_gtits1_reverse$CA$v <- -(test.pca_gtits1_reverse$CA$v)
```

```
# Figure
```

```
par(mfrow=c(1,1), mar=c(4,4,2,2), bty="n")
```

```
plot(test.pca_gtits1_reverse, scaling = scl, choices=c(1,2), type="n", ylim=c(-3,3), xlim=c(-3.5,3.5), main="", xlab="PC1 = 57%", ylab="PC2 = 43%", cex.lab=2, cex.axis=2)
```

```
with(data_gtits, points(test.pca_gtits1_reverse, display = "sites", col = col_points, scaling = scl, pch = 16, cex=2))
```

```
g <- vegan::scores(test.pca_gtits1_reverse, display = "species")
```

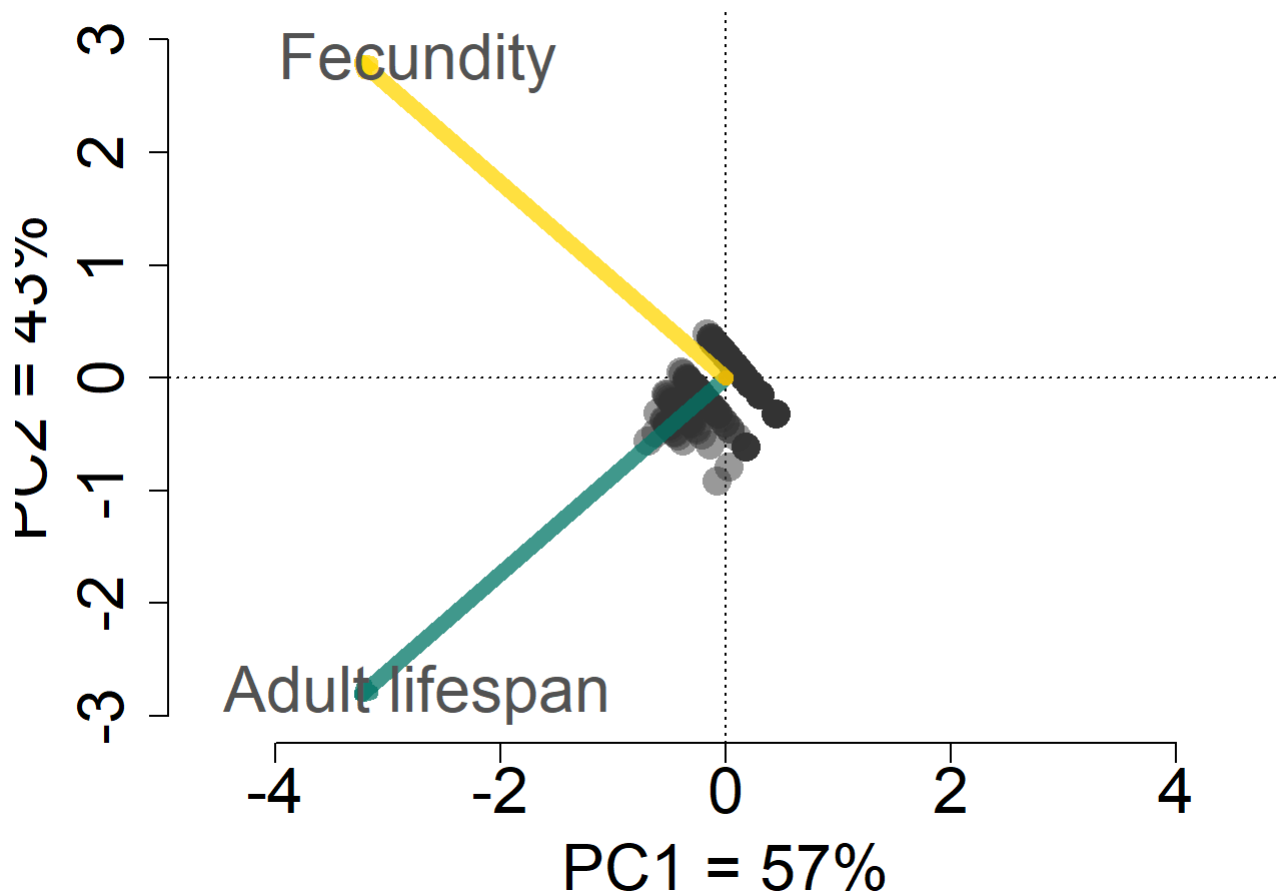
```
len <- 1
```

```
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, col_fec), lwd=8)
```

```
## Warning in arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col =
## c(col_afr, : zero-length arrow is of indeterminate angle and so skipped

## Warning in arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col =
## c(col_afr, : zero-length arrow is of indeterminate angle and so skipped
```

```
#text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32")
text(x=g[2,1]+0.5, y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
#text(x=g[3,1], y=g[3,2], labels="Breeding frequency", col = "gray32")
text(x=g[4,1]+0.5, y=g[4,2], labels="Fecundity", col = "gray32", cex=2)
```



```
# Test for proportion of variance explained per axe
sum_all_gtits <- summary(test.pca_gtits1)

prop_gtits <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:2){
    prop_gtits[j,i+1] <- as.numeric(abs(sum_all_gtits$species[i,j])/sum(abs(sum_all_gtits$species[i,])))
  }
}

prop_gtits
```

```
##   PC prim    adult breed    freq
## 1  1  NaN 0.5356924   NaN 0.5356924
## 2  2  NaN 0.4643076   NaN 0.4643076
## 3  3    3 3.0000000    3 3.0000000
## 4  4    4 4.0000000    4 4.0000000
```

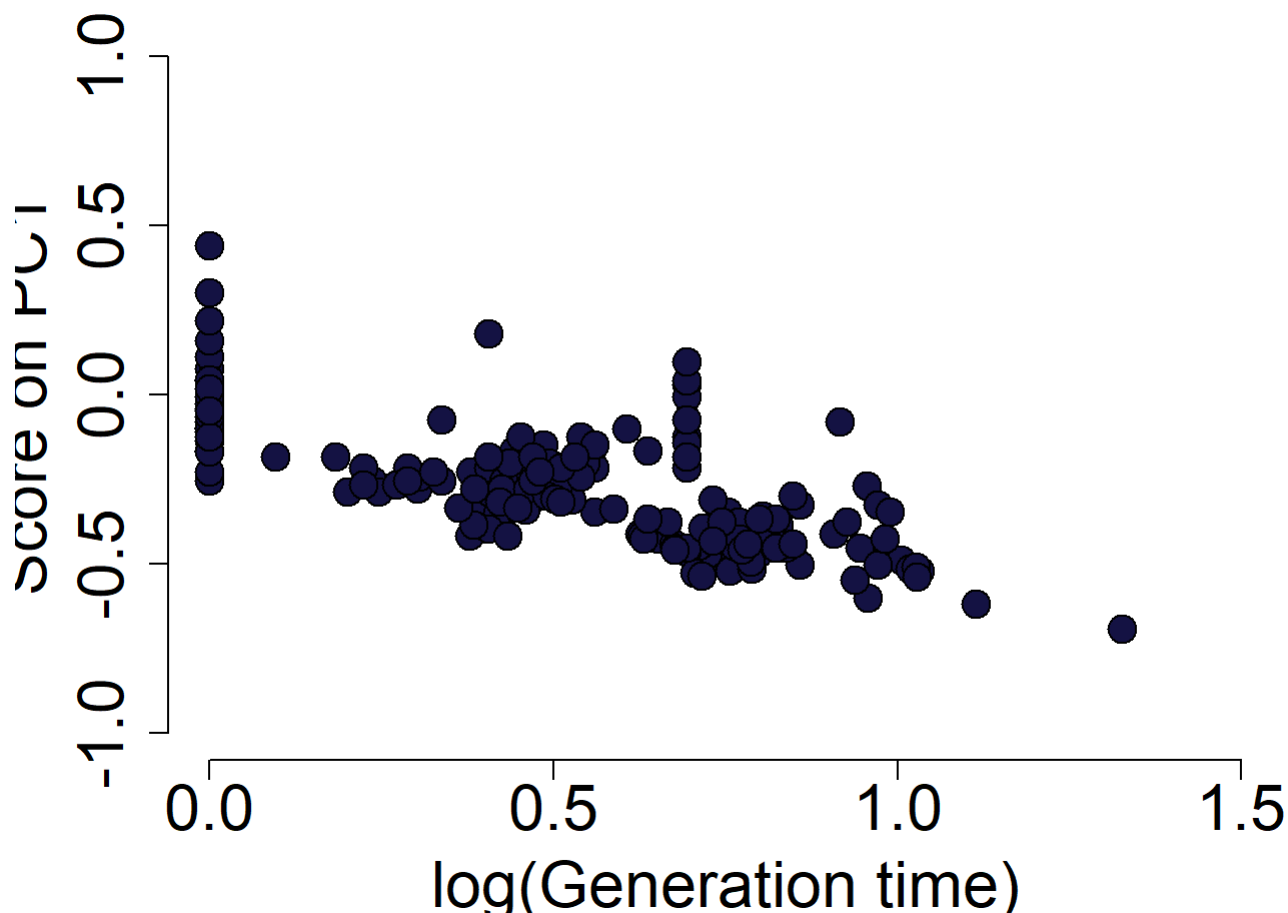
```
# Covariances
test <- as.matrix(scale(log(data_gtits[,c(2,3,4,7)])))
cov(test)
```

```
##           age_first_repro lifetime_adult breed_freq fec_one
## age_first_repro           NA           NA           NA      NA
## lifetime_adult           NA      1.0000000           NA 0.1420458
## breed_freq           NA           NA           NA      NA
## fec_one           NA      0.1420458           NA 1.0000000
```

```
# Correlation with generation time
individual_score_gtits_reverse <- as.numeric(vegan::scores(test.pca_gtits1_reverse)$sites[,1])
summary(lm(individual_score_gtits_reverse ~ log(data_gtits$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_gtits_reverse ~ log(data_gtits$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.34037 -0.11473 -0.04245  0.05514  0.43580
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.087310   0.006984   12.50  <2e-16 ***
## log(data_gtits$generation_time) -0.611347   0.022383  -27.31  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1608 on 669 degrees of freedom
## Multiple R-squared:  0.5272, Adjusted R-squared:  0.5265
## F-statistic: 746 on 1 and 669 DF, p-value: < 2.2e-16
```

```
par(bty="n")
plot(log(data_gtits$generation_time), individual_score_gtits_reverse, pch=21, xlab="log(Generati
on time)", ylab="Score on PC1", xlim=c(0,1.5), ylim=c(-1,1), bg=colors[7], cex=2, cex.axis=2, ce
x.lab=2)
```



Eurasian oystercatcher

```
data_oystercatcher <- data[which(data$species == "oystercatcher"),]  
data_oystercatcher$fec_one <- data_oystercatcher$fecundity + 1  
test.pca_oystercatcher1 <- rda(log(data_oystercatcher[,c(2,3,4,7)]), scale=TRUE)  
summary(test.pca_oystercatcher1)
```

```
##
## Call:
## rda(X = log(data_oystercatcher[, c(2, 3, 4, 7)]), scale = TRUE)
##
## Partitioning of correlations:
##           Inertia Proportion
## Total           4           1
## Unconstrained   4           1
##
## Eigenvalues, and their contribution to the correlations
##
## Importance of components:
##           PC1    PC2    PC3    PC4
## Eigenvalue      1.1768 1.1266 0.8853 0.8113
## Proportion Explained 0.2942 0.2817 0.2213 0.2028
## Cumulative Proportion 0.2942 0.5758 0.7972 1.0000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 8.122173
##
##
## Species scores
##
##           PC1    PC2    PC3    PC4
## age_first_repro 0.6063 -2.9670 2.5279 -0.9652
## lifetime_adult -2.9144 0.7241 1.9794 1.8858
## breed_freq      3.1179 0.3743 0.8021 2.4470
## fec_one         0.9084 3.0189 1.9102 -1.7043
##
##
## Site scores (weighted sums of species scores)
##
##           PC1    PC2    PC3    PC4
## 1306  0.1260950 -0.7963206 0.7408166 -0.159467
## 2108 -0.2236983 0.1398403 0.1322484 0.193718
## 3107 -0.1269198 0.0073315 -0.0110850 0.286447
## 4108  0.2958442 -0.4512319 0.0706535 -0.208867
## 5107  0.2239395 -0.0837178 -0.3278356 -0.042846
## 6108 -0.0991255 0.0001188 -0.0361773 0.260361
## 7107  0.0295787 0.1741290 0.0532261 0.040457
## 8107 -0.1362083 -0.4166952 0.5637615 0.174715
## 9106  0.0989609 -0.0512854 -0.2150067 0.074451
## 10107 -0.1280996 -0.3093215 0.6649617 0.114790
## 11107 -0.1730470 -0.4938505 0.6231947 0.085027
## 12103  0.0038255 -0.1713851 0.0632100 0.122842
## 1320  -0.4579245 -0.7824400 0.6390139 -0.560269
## 1419 -1.4305361 -0.2042153 -0.6124245 -1.457106
## 1520 -0.3524679 -0.2891698 0.2173096 -0.102979
## 1619 -0.1135577 0.0038640 -0.0231481 0.273906
## 1720 -0.0459566 -0.0333806 0.3612056 -0.319856
```

## 1820	0.3908970	-0.9370591	0.5974274	-0.428335
## 1919	0.4506659	-0.6361965	0.1232126	-0.395068
## 2019	-0.0019412	-0.0251009	-0.1239138	0.169150
## 2119	-0.1667581	0.1417856	0.1479109	0.264528
## 2219	-0.0779040	0.1276223	0.0765117	0.176885
## 2319	-0.0480608	-0.0249098	0.2418996	0.107980
## 2419	0.2781043	0.3770328	0.0939196	-0.320568
## 2519	-0.0936503	0.1228139	0.0819103	0.195914
## 2618	-0.3040587	-0.0200207	-0.0882903	0.057047
## 2718	0.2239395	-0.0837178	-0.3278356	-0.042846
## 2818	-0.5928702	0.0107463	-0.1518300	-0.385370
## 2918	0.2239395	-0.0837178	-0.3278356	-0.042846
## 3018	0.2239395	-0.0837178	-0.3278356	-0.042846
## 3118	0.0989609	-0.0512854	-0.2150067	0.074451
## 3218	-0.0260177	-0.0188530	-0.1021779	0.191747
## 3318	0.0279020	-0.1776330	0.0414741	0.100245
## 3418	-0.1269198	0.0073315	-0.0110850	0.286447
## 3518	-0.0662518	-0.0084121	-0.0658552	0.229508
## 3617	0.2781043	0.3770328	0.0939196	-0.320568
## 3717	-0.2088439	0.2687299	0.2607292	0.143341
## 3818	-0.0779040	0.1276223	0.0765117	0.176885
## 3917	0.0903666	0.5352295	0.3719271	-0.196684
## 4018	-0.0936503	0.1228139	0.0819103	0.195914
## 4118	-0.4341499	0.1706520	0.0438300	-0.264771
## 4218	-0.1726104	0.1583731	0.1002162	0.084947
## 4318	-0.5228181	0.0932824	0.0464986	-0.075649
## 4418	0.3010049	0.6500339	0.3497498	-0.475355
## 4518	0.1806748	-0.2172782	-0.0964470	-0.043137
## 4617	-0.3938225	-0.0213444	-0.0868686	0.008451
## 4717	-0.0019412	-0.0251009	-0.1239138	0.169150
## 4817	0.0587268	-0.0408445	-0.1786840	0.112212
## 4917	0.1930141	-0.9117697	1.1430500	-0.358914
## 5017	0.0699635	-0.3926150	0.2745752	0.003129
## 5117	0.4738140	-0.7545096	0.2514974	-0.448515
## 5217	-0.0032321	0.4366666	0.3346488	-0.050134
## 5317	-0.0991255	0.0001188	-0.0361773	0.260361
## 5417	-0.1191643	-0.4154788	0.6535888	-0.042435
## 5517	0.3256873	-0.6037640	0.2360414	-0.277772
## 5617	-0.2908716	-0.1928482	0.0125187	-0.129068
## 5717	0.0059284	0.4186147	0.3108412	-0.051241
## 5817	-0.0966848	-0.3081321	0.6458608	0.080843
## 5917	-0.1585305	-0.1671604	0.4327133	0.207945
## 6017	0.2383897	0.4326889	0.1747268	-0.304965
## 6116	-0.0934366	-0.0861574	0.2104800	0.185460
## 6216	0.3256873	-0.6037640	0.2360414	-0.277772
## 6316	-0.0692823	-0.1524133	0.1292106	0.191456
## 6416	-0.1777277	-0.0471407	-0.1840244	-0.030958
## 6515	0.0556963	-0.1848458	0.0163818	0.074160
## 6615	-0.0467195	0.2358676	0.1636791	0.092025
## 6715	0.2387274	-0.5139925	0.2252751	-0.177175
## 6815	0.0556963	-0.1848458	0.0163818	0.074160
## 6915	0.0575974	0.0307091	0.7021803	-0.186017

## 7014	-0.3201705	0.0175531	0.0342496	0.252362
## 7114	-0.0557331	-0.2599952	0.6671404	0.014328
## 7214	0.0998066	-0.5451471	0.4399631	-0.065775
## 7314	0.0556963	-0.1848458	0.0163818	0.074160
## 7414	-0.0579612	0.1968459	0.1322558	0.122616
## 7514	-0.0551884	-0.1086440	0.4240054	0.070178
## 7614	-0.0939340	0.1682374	0.1271198	0.174509
## 7714	-0.0972975	-0.3015336	0.4728355	0.137327
## 7814	0.0210710	-0.0960784	0.6000812	-0.086621
## 7914	-0.0985045	-0.4727225	0.3801699	-0.110207
## 8014	0.0392641	0.4219575	0.2926349	-0.088259
## 8114	-0.3017078	-0.3997275	0.5274390	-0.076538
## 8213	0.1508317	-0.0647461	-0.2618349	0.025768
## 8314	0.2769308	-0.3545631	-0.0341628	-0.165198
## 8414	0.0989609	-0.0512854	-0.2150067	0.074451
## 8513	-0.3538940	0.0841423	0.0321039	0.020863
## 8614	0.1268751	0.3183460	0.1333722	-0.131838
## 8713	-0.0662518	-0.0084121	-0.0658552	0.229508
## 8814	0.2239395	-0.0837178	-0.3278356	-0.042846
## 8914	0.0642599	0.1010010	-0.0416509	0.038551
## 9014	0.0313283	0.0903814	-0.0309185	0.078617
## 9114	-0.1743157	0.0658671	-0.0384080	0.005176
## 9213	-0.0315595	0.4689384	0.3849252	-0.035456
## 9313	-0.0435291	0.1931007	0.1192267	0.109071
## 9412	-0.0571861	0.0871668	0.0230355	0.174203
## 9512	-0.7261742	0.0018979	-0.3119413	-0.861248
## 9612	-0.1509963	0.0135795	0.0106509	0.309043
## 9713	0.0606740	0.1922441	0.0511092	-0.001239
## 9813	0.1268751	0.3183460	0.1333722	-0.131838
## 9913	0.0642599	0.1010010	-0.0416509	0.038551
## 10013	0.2518537	0.2859136	0.0205434	-0.249134
## 10113	0.1254435	0.5861147	0.3997234	-0.258270
## 10213	0.1507491	-0.6706730	0.5431557	-0.145308
## 10313	0.1268751	0.3183460	0.1333722	-0.131838
## 10413	0.1064464	0.5910446	0.4168737	-0.240441
## 10513	-0.1055707	0.1712572	0.1376252	0.185431
## 10613	0.0606740	0.1922441	0.0511092	-0.001239
## 10713	-0.1374123	0.2436483	0.2299386	0.184671
## 10813	-0.3689434	0.0514064	0.0681558	0.206460
## 10913	-0.0774465	0.0753691	0.0244201	0.201368
## 11013	0.1268751	0.3183460	0.1333722	-0.131838
## 11113	-0.1038339	0.2928818	0.2570650	0.125467
## 11213	0.0989609	-0.0512854	-0.2150067	0.074451
## 11312	-0.0467195	0.2358676	0.1636791	0.092025
## 11413	-0.0779040	0.1276223	0.0765117	0.176885
## 11513	-0.1338265	0.1524051	0.1371786	0.224462
## 11613	0.0258531	-0.0323136	-0.1490061	0.143065
## 11713	-0.0991255	0.0001188	-0.0361773	0.260361
## 11813	-0.0467195	0.2358676	0.1636791	0.092025
## 11913	-0.1540717	-0.2684000	0.4918380	0.049578
## 12013	-0.0274492	0.2489158	0.1641733	0.065315
## 12113	0.1339955	-0.0356989	0.1136782	-0.080306

##	12213	0.0203513	-0.5094085	0.6641863	-0.043499
##	12313	0.0510501	-0.2959462	0.1697589	0.046799
##	12412	-0.1089360	-0.2544291	0.3141920	0.196951
##	12513	0.2225079	0.1840509	-0.0614843	-0.169278
##	12613	0.3102600	-1.1312621	0.9559930	-0.413419
##	12713	0.4628843	-0.6986463	0.1909259	-0.423279
##	12813	0.1868566	-0.5005318	0.2721033	-0.128492
##	12911	-0.1733854	-0.0200604	0.3276099	0.238825
##	1307	-0.1412031	-0.5416633	0.1599226	-0.501943
##	13110	-0.1135577	0.0038640	-0.0231481	0.273906
##	1324	0.1868566	-0.5005318	0.2721033	-0.128492
##	1334	0.2227364	-0.4322602	0.1366541	-0.140253
##	1344	0.2387274	-0.5139925	0.2252751	-0.177175
##	1354	0.2387274	-0.5139925	0.2252751	-0.177175
##	1364	-0.2290989	-0.0924169	-0.3271874	-0.299000
##	1374	-0.0008835	0.2184892	0.1168633	0.051627
##	1384	0.1117181	-0.3116897	0.1149888	-0.010140
##	1394	0.1361646	-0.2665131	0.3568382	-0.127469
##	1404	-0.0233511	0.2046124	0.1176118	0.082130
##	14110	0.1425591	0.2501478	0.0556456	-0.115914
##	1424	-0.3027063	-0.0147083	-0.0711956	0.086338
##	1434	-0.2028803	-0.0462097	0.3207297	0.282659
##	1444	-0.1773619	0.1049787	0.1182711	0.293383
##	1454	-0.1860903	0.2254381	0.1870463	0.097734
##	1464	0.0989609	-0.0512854	-0.2150067	0.074451
##	1474	0.1466225	-0.4900909	0.3084260	-0.090731
##	1484	0.0258531	-0.0323136	-0.1490061	0.143065
##	1494	0.1425591	0.2501478	0.0556456	-0.115914
##	1504	-0.6792307	0.1276646	0.0464946	-0.279606
##	15110	-0.0208262	0.1492657	0.0611191	0.105895
##	1524	0.2383897	0.4326889	0.1747268	-0.304965
##	1534	-0.1871357	0.2401002	0.2585205	0.239200
##	1544	-0.0258978	0.3579912	0.2712933	0.011544
##	1554	0.0989609	-0.0512854	-0.2150067	0.074451
##	1564	0.1425591	0.2501478	0.0556456	-0.115914
##	1574	0.0244216	0.2354551	0.1173451	0.016632
##	1584	0.1999829	0.2993742	0.0673716	-0.200452
##	1593	-0.1191106	0.1650379	0.1402010	0.202789
##	1604	0.0989609	-0.0512854	-0.2150067	0.074451
##	16110	-0.0999555	0.4049574	0.3656568	0.067791
##	1624	0.3235299	0.5347106	0.2208940	-0.444181
##	1633	-0.3627077	0.0846799	0.0802213	0.139695
##	1644	-0.0180685	0.2229488	0.1323777	0.067756
##	1654	0.0989609	-0.0512854	-0.2150067	0.074451
##	1664	0.0587268	-0.0408445	-0.1786840	0.112212
##	1674	0.1041524	0.1168332	-0.0517097	-0.011401
##	1684	-0.1038339	0.2928818	0.2570650	0.125467
##	1694	-0.1269198	0.0073315	-0.0110850	0.286447
##	1704	-0.0190075	0.3320592	0.2411400	0.016615
##	17110	0.1999829	0.2993742	0.0673716	-0.200452
##	1724	0.2424530	-0.3075499	0.4275073	-0.279782
##	1733	0.1545573	0.1416965	-0.0596028	-0.076839

## 1744	0.3249615	0.2669418	-0.0454573	-0.317748
## 1754	0.0606740	0.1922441	0.0511092	-0.001239
## 1764	0.3010049	0.6500339	0.3497498	-0.475355
## 1774	0.0989609	-0.0512854	-0.2150067	0.074451
## 1784	0.2225079	0.1840509	-0.0614843	-0.169278
## 1794	0.0533253	0.2565564	0.1196030	-0.024162
## 1803	-0.3521011	0.0319886	0.0829636	0.315513
## 18110	-0.1102846	0.3206150	0.2099276	-0.087292
## 1824	-0.1055707	0.1712572	0.1376252	0.185431
## 1834	0.3966377	0.5157388	0.1548933	-0.512795
## 1844	-0.0019412	-0.0251009	-0.1239138	0.169150
## 1854	-0.0208262	0.1492657	0.0611191	0.105895
## 1864	-0.0260177	-0.0188530	-0.1021779	0.191747
## 1874	0.0587268	-0.0408445	-0.1786840	0.112212
## 1883	-0.0617578	0.0712978	0.0102566	0.186644
## 1893	-0.5237959	-0.1138800	-0.3778787	-0.550910
## 1903	-0.2069343	0.1713769	0.2031792	0.293076
## 19110	-0.0274492	0.2489158	0.1641733	0.065315
## 1923	-0.1692021	0.0659805	0.0743465	0.303348
## 1933	-0.2115218	0.1036849	0.1390406	0.330297
## 1943	0.0989609	-0.0512854	-0.2150067	0.074451
## 1953	0.0587268	-0.0408445	-0.1786840	0.112212
## 1963	0.0977578	-0.3998277	0.2494830	-0.022956
## 1973	-0.0740779	-0.2573043	0.5016887	0.091520
## 1983	-0.0662518	-0.0084121	-0.0658552	0.229508
## 1993	-0.5759319	-0.0541076	-0.1812579	-0.252443
## 2003	0.0989609	-0.0512854	-0.2150067	0.074451
## 20110	0.1545573	0.1416965	-0.0596028	-0.076839
## 2023	-0.1868515	0.1089406	0.0604094	0.123700
## 2033	-0.1794231	0.1868932	0.0490574	-0.140990
## 2043	-0.1200881	-0.0611810	0.5129493	0.116457
## 2053	0.2038230	-0.3355913	0.0318378	-0.096584
## 2063	0.0587268	-0.0408445	-0.1786840	0.112212
## 2073	0.0587268	-0.0408445	-0.1786840	0.112212
## 2083	-0.0719360	0.1483379	0.0931934	0.160645
## 2093	-0.0779040	0.1276223	0.0765117	0.176885
## 2109	-0.1470363	-0.2445420	0.3485883	0.232709
## 21110	-0.1005570	0.2678875	0.2301739	0.133929
## 2123	-0.1575737	0.0319328	0.3126757	0.049817
## 2133	0.0510501	-0.2959462	0.1697589	0.046799
## 2143	0.0989609	-0.0512854	-0.2150067	0.074451
## 2153	-0.3705375	0.1240817	0.1210464	0.120578
## 2163	0.2781043	0.3770328	0.0939196	-0.320568
## 2173	-0.0233511	0.2046124	0.1176118	0.082130
## 2183	-0.0991255	0.0001188	-0.0361773	0.260361
## 2193	-0.1237766	0.2236582	0.2013208	0.179735
## 2203	-0.1216505	0.1154421	0.0926786	0.229187
## 22110	0.0642599	0.1010010	-0.0416509	0.038551
## 2223	0.2262335	0.3904935	0.1407478	-0.271886
## 2233	-0.2106899	0.1065725	0.1413659	0.328033
## 2243	0.1545573	0.1416965	-0.0596028	-0.076839
## 2253	-0.0791828	0.0693424	0.0195669	0.206093

##	2263	-0.1005570	0.2678875	0.2301739	0.133929
##	2273	-0.1957362	-0.0727486	-0.2656980	-0.177459
##	2283	0.0989609	-0.0512854	-0.2150067	0.074451
##	2293	-0.0991255	0.0001188	-0.0361773	0.260361
##	2303	0.1508317	-0.0647461	-0.2618349	0.025768
##	23110	0.2239395	-0.0837178	-0.3278356	-0.042846
##	2323	-0.1509963	0.0135795	0.0106509	0.309043
##	2333	0.0989609	-0.0512854	-0.2150067	0.074451
##	2343	-0.0019412	-0.0251009	-0.1239138	0.169150
##	2353	0.3488354	-0.7220771	0.3643262	-0.331219
##	2363	0.1254435	0.5861147	0.3997234	-0.258270
##	2373	-0.1269198	0.0073315	-0.0110850	0.286447
##	2383	0.0989609	-0.0512854	-0.2150067	0.074451
##	2393	0.0536552	0.1678810	0.0314901	0.017861
##	2403	0.2239395	-0.0837178	-0.3278356	-0.042846
##	24110	0.0975293	0.2164833	0.0513445	-0.051982
##	2423	0.0313283	0.0903814	-0.0309185	0.078617
##	2433	0.3489181	-0.1161503	-0.4406644	-0.160142
##	2443	0.1508317	-0.0647461	-0.2618349	0.025768
##	2453	0.1508317	-0.0647461	-0.2618349	0.025768
##	2463	0.1758286	0.3656302	0.1486409	-0.206448
##	2473	0.0258531	-0.0323136	-0.1490061	0.143065
##	2483	0.3489181	-0.1161503	-0.4406644	-0.160142
##	2493	0.1519522	-0.3221306	0.0786660	-0.047901
##	2503	0.0989609	-0.0512854	-0.2150067	0.074451
##	25110	-0.2283217	-0.0819730	0.4390016	-0.081292
##	2523	0.2038230	-0.3355913	0.0318378	-0.096584
##	2532	0.0989609	-0.0512854	-0.2150067	0.074451
##	2542	0.1508317	-0.0647461	-0.2618349	0.025768
##	2552	-0.1317776	0.0070858	0.3276588	0.181643
##	2562	-0.3891477	0.0432933	0.0027206	0.047331
##	2572	0.1508317	-0.0647461	-0.2618349	0.025768
##	2582	-0.0949725	-0.2934636	0.4793342	0.131000
##	2592	0.2337440	-0.7401792	0.5319487	-0.236751
##	2602	-0.1135577	0.0038640	-0.0231481	0.273906
##	2619	0.3489181	-0.1161503	-0.4406644	-0.160142
##	2622	-0.2364560	0.0257948	0.2978363	-0.019732
##	2632	-1.1310829	-0.3786173	-0.0225048	-0.945217
##	2642	-0.5991382	-0.2132909	-0.0340103	-0.381737
##	2652	-0.0472547	-0.0133419	-0.0830055	0.211679
##	2662	-0.0991255	0.0001188	-0.0361773	0.260361
##	2672	0.0989609	-0.0512854	-0.2150067	0.074451
##	2682	-1.1463343	0.0487949	-0.2774078	-1.178903
##	2692	0.3966377	0.5157388	0.1548933	-0.512795
##	2702	0.5216163	0.4833064	0.0420645	-0.630091
##	2719	-0.1135577	0.0038640	-0.0231481	0.273906
##	2722	0.3489181	-0.1161503	-0.4406644	-0.160142
##	2732	-0.2078129	-0.0135082	-0.0733432	0.136635
##	2742	-0.5364752	-0.0126515	-0.0496926	-0.010517
##	2752	-0.1221700	0.0702270	0.0481942	0.251345
##	2762	-0.3936401	-0.0086690	-0.0458872	0.076934
##	2772	-0.0260177	-0.0188530	-0.1021779	0.191747

##	2782	-0.3540775	-0.0599845	-0.2143586	-0.181703
##	2792	0.1508317	-0.0647461	-0.2618349	0.025768
##	2802	-0.0575038	0.1445927	0.0801643	0.147100
##	2819	-0.2048398	0.0707919	0.1021217	0.338915
##	2822	0.0989609	-0.0512854	-0.2150067	0.074451
##	2832	-0.1135577	0.0038640	-0.0231481	0.273906
##	2841	-0.4163902	-0.0154880	-0.0664949	0.029632
##	2852	-0.3088187	0.0057178	-0.0047478	0.193738
##	2862	-0.0954934	-0.5654547	0.8064158	0.066055
##	2872	0.0989609	-0.0512854	-0.2150067	0.074451
##	2882	0.0989609	-0.0512854	-0.2150067	0.074451
##	2892	-0.0936149	0.0730876	0.0325960	0.219638
##	2902	-0.0834368	-0.0039525	-0.0503408	0.245637
##	2919	0.2225079	0.1840509	-0.0614843	-0.169278
##	2922	-0.1629898	0.1861577	0.1894623	0.239320
##	2932	-0.0118098	-0.5161822	0.5407288	0.038980
##	2942	-0.7741719	0.2296936	0.0762927	-0.574840
##	2952	-0.0019412	-0.0251009	-0.1239138	0.169150
##	2962	-0.6401020	-0.1051123	-0.3421341	-0.557307
##	2972	0.2239395	-0.0837178	-0.3278356	-0.042846
##	2982	-0.0774465	0.0753691	0.0244201	0.201368
##	2992	0.0313283	0.0903814	-0.0309185	0.078617
##	3001	0.2225079	0.1840509	-0.0614843	-0.169278
##	3019	-0.0190075	0.3320592	0.2411400	0.016615
##	3022	-0.2308580	0.0582502	0.0532631	0.233161
##	3032	-0.1864002	-0.0234760	-0.1069401	0.092736
##	3042	-0.0939334	0.0711261	0.3667087	0.110830
##	3052	-0.6938303	-0.0077209	-0.1906130	-0.493598
##	3062	0.0258531	-0.0323136	-0.1490061	0.143065
##	3072	0.1508317	-0.0647461	-0.2618349	0.025768
##	3081	-0.0472547	-0.0133419	-0.0830055	0.211679
##	3092	-0.0779040	0.1276223	0.0765117	0.176885
##	3108	-0.1799039	0.1573474	0.1718229	0.271059
##	3119	0.0989609	-0.0512854	-0.2150067	0.074451
##	3122	-0.0123446	0.3013459	0.2063941	0.024211
##	3132	-0.5694267	-0.0307088	-0.1059977	-0.123174
##	3142	0.0027426	0.1602048	0.0567477	0.075625
##	3152	-0.0514639	-0.4386867	0.4872555	0.095179
##	3162	-0.5988604	-0.0685753	-0.2265909	-0.341101
##	3172	-0.0617578	0.0712978	0.0102566	0.186644
##	3182	0.1545573	0.1416965	-0.0596028	-0.076839
##	3192	-0.1416590	0.2901002	0.2787258	0.166987
##	3202	-0.2585201	0.0189696	0.0349140	0.288475
##	3219	-0.2084154	0.0284799	0.0624880	0.362933
##	3222	0.0989609	-0.0512854	-0.2150067	0.074451
##	3232	-0.0739285	-0.2635137	0.2825877	0.164095
##	3242	-0.1096645	-0.3976961	0.2695545	-0.107241
##	3252	-0.1089360	-0.2544291	0.3141920	0.196951
##	3262	0.0989609	-0.0512854	-0.2150067	0.074451
##	3272	-0.1819819	0.0216203	0.0386243	0.338124
##	3282	0.2239395	-0.0837178	-0.3278356	-0.042846
##	3292	0.4628843	-0.6986463	0.1909259	-0.423279

##	3302	-0.0662518	-0.0084121	-0.0658552	0.229508
##	3319	-0.8164518	-0.1179560	-0.3724683	-0.708052
##	3322	-0.3221463	0.2274526	0.1134820	-0.186595
##	3332	0.0628283	0.3687697	0.2247003	-0.087881
##	3342	0.0628283	0.3687697	0.2247003	-0.087881
##	3351	0.3787612	-0.2686824	-0.2752765	-0.229047
##	3362	0.3010049	0.6500339	0.3497498	-0.475355
##	3372	0.0587268	-0.0408445	-0.1786840	0.112212
##	3382	-0.3115426	-0.1227176	-0.4199447	-0.500581
##	3392	-0.0508337	0.2714357	0.2015921	0.079400
##	3402	0.1120250	-0.6075970	0.5076764	-0.093987
##	3419	-0.0785456	0.1752441	0.1241312	0.154810
##	3422	0.1545573	0.1416965	-0.0596028	-0.076839
##	3432	0.2647979	-0.6472421	0.3697553	-0.237369
##	3442	0.0313283	0.0903814	-0.0309185	0.078617
##	3452	0.3040555	-0.6796078	0.9840327	-0.483782
##	3462	-0.0269668	-0.5268762	0.9453539	-0.272831
##	3472	-0.1319846	0.1951670	0.1783775	0.202071
##	3482	-0.4416174	-0.0372675	-0.1353288	-0.099544
##	3492	0.5010308	-0.8936179	0.4023302	-0.511356
##	3502	0.2239395	-0.0837178	-0.3278356	-0.042846
##	3519	-0.0008835	0.2184892	0.1168633	0.051627
##	3522	-0.0596902	0.0784749	0.0160361	0.181017
##	3532	0.3489181	-0.1161503	-0.4406644	-0.160142
##	3542	0.1508317	-0.0647461	-0.2618349	0.025768
##	3552	-0.8522676	-0.0624187	-0.1905812	-0.424902
##	3562	-0.1777277	-0.0471407	-0.1840244	-0.030958
##	3572	0.0989609	-0.0512854	-0.2150067	0.074451
##	3582	-0.0953999	0.2065614	0.1660549	0.157754
##	3592	0.2239395	-0.0837178	-0.3278356	-0.042846
##	3602	-0.0019412	-0.0251009	-0.1239138	0.169150
##	3618	0.0587268	-0.0408445	-0.1786840	0.112212
##	3622	-0.1819819	0.0216203	0.0386243	0.338124
##	3632	-0.0400011	0.0827072	0.0075211	0.158074
##	3642	-0.1863707	0.0565109	-0.0027495	0.119755
##	3652	0.1508317	-0.0647461	-0.2618349	0.025768
##	3662	0.0587268	-0.0408445	-0.1786840	0.112212
##	3672	-0.0662518	-0.0084121	-0.0658552	0.229508
##	3682	-0.1468226	0.0688463	0.0627401	0.278199
##	3692	-0.1211531	-0.1389527	0.1760388	0.240139
##	3702	-0.0662518	-0.0084121	-0.0658552	0.229508
##	3718	0.3489181	-0.1161503	-0.4406644	-0.160142
##	3722	-0.9508601	0.0828603	-0.2383834	-1.058619
##	3732	0.0153674	0.3975376	0.2838551	-0.051199
##	3742	0.3489181	-0.1161503	-0.4406644	-0.160142
##	3752	0.0989609	-0.0512854	-0.2150067	0.074451
##	3762	0.0989609	-0.0512854	-0.2150067	0.074451
##	3772	-1.7624373	-0.2283881	-0.6695154	-1.740818
##	3782	-0.0162545	0.1651346	0.0738980	0.093455
##	3792	-0.0019412	-0.0251009	-0.1239138	0.169150
##	3802	0.0587268	-0.0408445	-0.1786840	0.112212
##	3819	0.0258531	-0.0323136	-0.1490061	0.143065

##	3822	-0.0707139	0.1153554	0.3955618	0.065024
##	3832	-0.1269198	0.0073315	-0.0110850	0.286447
##	3842	-0.3223232	-0.3253186	0.0984866	-0.284124
##	3852	0.0989609	-0.0512854	-0.2150067	0.074451
##	3862	0.1806748	-0.2172782	-0.0964470	-0.043137
##	3872	0.0556963	-0.1848458	0.0163818	0.074160
##	3882	0.1041524	0.1168332	-0.0517097	-0.011401
##	3892	-0.4687739	-0.5545956	0.1389092	-0.722999
##	3902	-0.1269198	0.0073315	-0.0110850	0.286447
##	3918	-0.1097301	0.0669988	0.0369637	0.239670
##	3922	-0.1135577	0.0038640	-0.0231481	0.273906
##	3932	0.1508317	-0.0647461	-0.2618349	0.025768
##	3942	0.0244216	0.2354551	0.1173451	0.016632
##	3952	0.1087654	-0.7077468	0.6447775	-0.119454
##	3962	-0.4097156	-0.0669603	-0.2333841	-0.245040
##	3972	0.5216163	0.4833064	0.0420645	-0.630091
##	3982	0.6226383	0.8339660	0.3244428	-0.904994
##	3992	0.0556963	-0.1848458	0.0163818	0.074160
##	4002	0.2225079	0.1840509	-0.0614843	-0.169278
##	4019	0.0279020	-0.1776330	0.0414741	0.100245
##	4022	0.3489181	-0.1161503	-0.4406644	-0.160142
##	4032	-0.0472547	-0.0133419	-0.0830055	0.211679
##	4042	0.3249615	0.2669418	-0.0454573	-0.317748
##	4052	-0.0472547	-0.0133419	-0.0830055	0.211679
##	4062	-0.0662518	-0.0084121	-0.0658552	0.229508
##	4072	0.2239395	-0.0837178	-0.3278356	-0.042846
##	4082	-0.3540775	-0.0599845	-0.2143586	-0.181703
##	4092	0.3489181	-0.1161503	-0.4406644	-0.160142
##	4109	-0.0461342	-0.2707265	0.2574954	0.138009
##	4119	-0.0036366	0.2089330	0.1091679	0.059119
##	4122	-0.3122727	-0.0326435	-0.1285917	-0.014861
##	4132	-0.0991255	0.0001188	-0.0361773	0.260361
##	4141	0.2239395	-0.0837178	-0.3278356	-0.042846
##	4152	-0.1619273	0.0164161	0.0205192	0.319303
##	4161	-0.0019412	-0.0251009	-0.1239138	0.169150
##	4172	0.1288040	-0.2038175	-0.0496188	0.005546
##	4182	0.1508317	-0.0647461	-0.2618349	0.025768
##	4192	-0.2983132	0.1205502	0.1297076	0.199489
##	4202	-0.1509963	0.0135795	0.0106509	0.309043
##	4219	-0.1135577	0.0038640	-0.0231481	0.273906
##	4222	-0.0472547	-0.0133419	-0.0830055	0.211679
##	4232	0.2537826	-0.2362500	-0.1624476	-0.111751
##	4242	-0.1298996	-0.4084068	0.2362006	-0.174383
##	4252	0.2537826	-0.2362500	-0.1624476	-0.111751
##	4262	-0.0418151	0.1405214	0.0660008	0.132375
##	4272	0.3787612	-0.2686824	-0.2752765	-0.229047
##	4282	0.1806748	-0.2172782	-0.0964470	-0.043137
##	4292	-0.2362800	0.0695142	0.1211515	0.372932
##	4302	-0.8566859	-0.1075151	-0.3361455	-0.670291
##	4319	-0.1658930	-0.2252807	-0.1003101	-0.246364
##	4322	-0.0991255	0.0001188	-0.0361773	0.260361
##	4332	0.2227364	-0.4322602	0.1366541	-0.140253

##	4342	0.0989609	-0.0512854	-0.2150067	0.074451
##	4352	0.3489181	-0.1161503	-0.4406644	-0.160142
##	4362	-0.3115426	-0.1227176	-0.4199447	-0.500581
##	4372	-0.0791828	0.0693424	0.0195669	0.206093
##	4382	0.2225079	0.1840509	-0.0614843	-0.169278
##	4392	0.1508317	-0.0647461	-0.2618349	0.025768
##	4402	0.2781043	0.3770328	0.0939196	-0.320568
##	4419	-0.0837145	-0.1486681	0.1422398	0.205001
##	4422	0.3489181	-0.1161503	-0.4406644	-0.160142
##	4432	0.2225079	0.1840509	-0.0614843	-0.169278
##	4442	0.0295787	0.1741290	0.0532261	0.040457
##	4452	0.1806748	-0.2172782	-0.0964470	-0.043137
##	4462	-0.1135577	0.0038640	-0.0231481	0.273906
##	4472	-0.0991255	0.0001188	-0.0361773	0.260361
##	4482	-0.0260177	-0.0188530	-0.1021779	0.191747
##	4492	0.1508317	-0.0647461	-0.2618349	0.025768
##	4502	-0.0472547	-0.0133419	-0.0830055	0.211679
##	4519	0.1508317	-0.0647461	-0.2618349	0.025768
##	4522	0.2239395	-0.0837178	-0.3278356	-0.042846
##	4532	0.3489181	-0.1161503	-0.4406644	-0.160142
##	4542	0.2518537	0.2859136	0.0205434	-0.249134
##	4552	0.0258531	-0.0323136	-0.1490061	0.143065
##	4562	0.0587268	-0.0408445	-0.1786840	0.112212
##	4572	0.0033281	0.0830097	-0.0201502	0.111891
##	4582	0.0587268	-0.0408445	-0.1786840	0.112212
##	4592	0.2225079	0.1840509	-0.0614843	-0.169278
##	4602	-0.1921243	-0.0175795	-0.0875067	0.121911
##	4618	0.1508317	-0.0647461	-0.2618349	0.025768
##	4622	0.6943145	1.0827630	0.5247934	-1.100040
##	4632	-0.1135577	0.0038640	-0.0231481	0.273906
##	4642	-0.0472547	-0.0133419	-0.0830055	0.211679
##	4652	0.2225079	0.1840509	-0.0614843	-0.169278
##	4662	-0.0791828	0.0693424	0.0195669	0.206093
##	4672	0.2225079	0.1840509	-0.0614843	-0.169278
##	4682	0.1281990	0.0348466	0.8406663	-0.321997
##	4692	-0.0753789	0.0825462	0.0301996	0.195741
##	4702	0.2239395	-0.0837178	-0.3278356	-0.042846
##	4718	0.0989609	-0.0512854	-0.2150067	0.074451
##	4722	-0.0728748	0.0912380	0.0371990	0.188927
##	4732	0.0258531	-0.0323136	-0.1490061	0.143065
##	4742	0.1508317	-0.0647461	-0.2618349	0.025768
##	4751	0.3489181	-0.1161503	-0.4406644	-0.160142
##	4762	-0.0991255	0.0001188	-0.0361773	0.260361
##	4772	-0.1119038	0.2182438	-0.0313685	-0.405585
##	4782	-0.6641785	-0.0988643	-0.3203982	-0.534710
##	4792	0.1174547	0.8555851	0.6719944	-0.378549
##	4802	0.3433354	0.7969682	0.4680727	-0.590545
##	4818	-0.1315719	0.1433647	0.0108186	-0.077273
##	4822	0.3952062	0.7835075	0.4212445	-0.639227
##	4832	-0.0106554	0.1845699	0.0895488	0.078218
##	4842	-0.1476862	0.0658486	0.0603261	0.280550
##	4852	0.0258531	-0.0323136	-0.1490061	0.143065

##	4862	-0.0791828	0.0693424	0.0195669	0.206093
##	4872	-0.0991255	0.0001188	-0.0361773	0.260361
##	4882	-0.1904899	0.4483203	0.2196965	-0.504655
##	4892	0.3489181	-0.1161503	-0.4406644	-0.160142
##	4902	-0.0853698	0.3569728	0.3086760	0.075222
##	4918	0.0989609	-0.0512854	-0.2150067	0.074451
##	4922	-0.2252924	0.3204581	0.2642877	-0.016684
##	4932	0.3966377	0.5157388	0.1548933	-0.512795
##	4942	-0.0991255	0.0001188	-0.0361773	0.260361
##	4952	-0.0472547	-0.0133419	-0.0830055	0.211679
##	4962	0.1508317	-0.0647461	-0.2618349	0.025768
##	4972	-0.0991255	0.0001188	-0.0361773	0.260361
##	4982	-0.0991255	0.0001188	-0.0361773	0.260361
##	4992	0.1508317	-0.0647461	-0.2618349	0.025768
##	5002	-0.0834368	-0.0039525	-0.0503408	0.245637
##	5018	-0.0607187	0.1334334	0.0711779	0.155848
##	5022	-0.0873070	0.0949833	0.0502281	0.202472
##	5032	0.1508317	-0.0647461	-0.2618349	0.025768
##	5042	-0.0991255	0.0001188	-0.0361773	0.260361
##	5052	-0.2000276	0.0263033	0.0549156	0.355061
##	5062	-0.4465872	-0.0494836	-0.1745207	-0.167760
##	5072	0.1508317	-0.0647461	-0.2618349	0.025768
##	5082	-0.1135577	0.0038640	-0.0231481	0.273906
##	5092	0.1508317	-0.0647461	-0.2618349	0.025768
##	5108	-0.1790743	0.0662157	0.0809526	0.313725
##	5118	-0.3070368	-0.0259487	-0.1072730	0.023683
##	5122	-0.0991255	0.0001188	-0.0361773	0.260361
##	5132	-0.0260177	-0.0188530	-0.1021779	0.191747
##	5142	-0.0472547	-0.0133419	-0.0830055	0.211679
##	5152	0.2781043	0.3770328	0.0939196	-0.320568
##	5162	0.2239395	-0.0837178	-0.3278356	-0.042846
##	5172	-0.0274492	0.2489158	0.1641733	0.065315
##	5182	0.1041524	0.1168332	-0.0517097	-0.011401
##	5192	-0.0300316	0.4334627	0.3487734	-0.020127
##	5202	-0.2536479	0.1235114	0.1858883	0.365583
##	5218	-0.0123446	0.3013459	0.2063941	0.024211
##	5222	0.0258531	-0.0323136	-0.1490061	0.143065
##	5232	-0.4036475	-0.0988161	-0.3367937	-0.414137
##	5242	-0.4365212	-0.0902852	-0.3071159	-0.383284
##	5252	0.0027426	0.1602048	0.0567477	0.075625
##	5262	-0.1382169	0.2078848	0.0912927	-0.069017
##	5272	-0.3022067	-0.0734451	-0.2611868	-0.230386
##	5282	-0.2035968	-0.1692534	0.0832815	0.038558
##	5292	0.0389223	-0.2604438	0.4256293	-0.027046
##	5302	0.3379057	-0.6662139	0.3037547	-0.305983
##	5318	0.1806748	-0.2172782	-0.0964470	-0.043137
##	5322	-0.1779698	-0.1660404	0.0920447	0.067731
##	5332	-0.3150293	-0.2988853	0.1835105	-0.138116
##	5342	0.4019094	-0.3869955	-0.1469916	-0.282494
##	5352	-0.1777277	-0.0471407	-0.1840244	-0.030958
##	5362	0.2007087	-0.5713316	0.3488702	-0.160475
##	5371	0.4019094	-0.3869955	-0.1469916	-0.282494

##	5382	-0.1779698	-0.1660404	0.0920447	0.067731
##	5392	0.0885699	-0.1933766	-0.0132961	0.043307
##	5402	-0.1009528	0.1438743	0.1075007	0.193609
##	5418	-0.3634134	-0.1092570	-0.3731165	-0.451898
##	5422	-0.1402905	-0.5504980	0.4379113	-0.205526
##	5432	0.0258531	-0.0323136	-0.1490061	0.143065
##	5442	-0.2849085	-0.3067018	0.1563179	-0.166386
##	5451	0.2129271	-0.6337814	0.4165835	-0.188686
##	5462	0.1120250	-0.6075970	0.5076764	-0.093987
##	5472	-0.1135577	0.0038640	-0.0231481	0.273906
##	5482	0.1508317	-0.0647461	-0.2618349	0.025768
##	5492	0.0989609	-0.0512854	-0.2150067	0.074451
##	5502	0.2239395	-0.0837178	-0.3278356	-0.042846
##	5518	0.1508317	-0.0647461	-0.2618349	0.025768
##	5522	-0.1135577	0.0038640	-0.0231481	0.273906
##	5532	0.2239395	-0.0837178	-0.3278356	-0.042846
##	5542	-0.1269198	0.0073315	-0.0110850	0.286447
##	5552	0.0989609	-0.0512854	-0.2150067	0.074451
##	5562	-0.4783704	0.0437977	-0.2509044	-0.674522
##	5572	0.0989609	-0.0512854	-0.2150067	0.074451
##	5582	0.4506659	-0.6361965	0.1232126	-0.395068
##	5592	0.1306315	-0.4083586	0.2198051	-0.053809
##	5602	0.2958442	-0.4512319	0.0706535	-0.208867
##	5618	-0.2222451	-0.0097630	-0.0603140	0.150181
##	5622	0.2239395	-0.0837178	-0.3278356	-0.042846
##	5632	0.0975293	0.2164833	0.0513445	-0.051982
##	5642	-0.0335646	0.3313788	0.2498629	0.032407
##	5652	0.0989609	-0.0512854	-0.2150067	0.074451
##	5662	-0.2823078	0.1837379	0.0665536	-0.150098
##	5672	0.1143232	0.1521374	-0.0232800	-0.039078
##	5682	-0.0534719	0.2124289	0.1448045	0.110400
##	5692	-0.0596902	0.0784749	0.0160361	0.181017
##	5702	-0.0390397	0.2086837	0.1317753	0.096855
##	5718	-0.0834368	-0.0039525	-0.0503408	0.245637
##	5722	-0.0202025	0.3279113	0.2377998	0.019867
##	5732	-0.1583319	0.2619367	0.1677505	-0.047562
##	5742	0.2239395	-0.0837178	-0.3278356	-0.042846
##	5752	0.0989609	-0.0512854	-0.2150067	0.074451
##	5762	-0.0918787	0.0791143	0.0374492	0.214913
##	5772	0.1868566	-0.5005318	0.2721033	-0.128492
##	5782	-0.0491282	0.1736654	0.1035759	0.124308
##	5792	0.3249615	0.2669418	-0.0454573	-0.317748
##	5802	0.0186488	0.4627689	0.3463976	-0.085856
##	5818	-0.1135577	0.0038640	-0.0231481	0.273906
##	5822	-0.0991255	0.0001188	-0.0361773	0.260361
##	5832	0.0269736	-0.2896982	0.1914948	0.069395
##	5842	0.0269736	-0.2896982	0.1914948	0.069395
##	5852	0.2537826	-0.2362500	-0.1624476	-0.111751
##	5862	0.0964757	-0.1382684	0.2967332	-0.076814
##	5872	0.0723992	-0.1320205	0.3184691	-0.054217
##	5882	-0.1135577	0.0038640	-0.0231481	0.273906
##	5892	-1.1440225	-0.1308883	-0.3934818	-0.929109

##	5902	0.0556963	-0.1848458	0.0163818	0.074160
##	5918	0.2781043	0.3770328	0.0939196	-0.320568
##	5922	0.0269736	-0.2896982	0.1914948	0.069395
##	5932	0.3489181	-0.1161503	-0.4406644	-0.160142
##	5942	0.3489181	-0.1161503	-0.4406644	-0.160142
##	5952	-0.0834368	-0.0039525	-0.0503408	0.245637
##	5962	-0.1620244	-0.4124346	0.2252155	-0.210952
##	5972	-0.0242479	-0.0546219	0.1970748	0.096876
##	5982	0.2239395	-0.0837178	-0.3278356	-0.042846
##	5992	0.4019094	-0.3869955	-0.1469916	-0.282494
##	6002	0.0885699	-0.1933766	-0.0132961	0.043307
##	6018	-0.1700499	-0.3062269	0.1505567	-0.110782
##	6022	-0.1886819	-0.1800359	0.0474626	-0.012739
##	6032	0.2769308	-0.3545631	-0.0341628	-0.165198
##	6042	0.0788444	-0.3031589	0.1446666	0.020713
##	6052	-0.0364087	-0.1609442	0.0995327	0.160603
##	6062	0.2537826	-0.2362500	-0.1624476	-0.111751
##	6072	-0.0070629	-0.0590815	0.1815605	0.080747
##	6082	0.1288040	-0.2038175	-0.0496188	0.005546
##	6092	0.2605887	-1.0725870	0.9400163	-0.353868
##	6109	-0.0834368	-0.0039525	-0.0503408	0.245637
##	6117	0.1508317	-0.0647461	-0.2618349	0.025768
##	6122	-0.0260177	-0.0188530	-0.1021779	0.191747
##	6132	0.2239395	-0.0837178	-0.3278356	-0.042846
##	6142	0.3489181	-0.1161503	-0.4406644	-0.160142
##	6152	-0.0472547	-0.0133419	-0.0830055	0.211679
##	6162	-0.2744046	0.1267495	-0.0806905	-0.355590
##	6172	-0.0274492	0.2489158	0.1641733	0.065315
##	6182	-0.0019412	-0.0251009	-0.1239138	0.169150
##	6192	-0.0834368	-0.0039525	-0.0503408	0.245637
##	6202	-0.0179089	0.0885208	-0.0009778	0.131823
##	6217	-0.0179089	0.0885208	-0.0009778	0.131823
##	6222	-0.0834368	-0.0039525	-0.0503408	0.245637
##	6232	-0.0662518	-0.0084121	-0.0658552	0.229508
##	6242	-0.0274492	0.2489158	0.1641733	0.065315
##	6252	0.5216163	0.4833064	0.0420645	-0.630091
##	6262	-0.0662518	-0.0084121	-0.0658552	0.229508
##	6272	0.0989609	-0.0512854	-0.2150067	0.074451
##	6282	-0.0019412	-0.0251009	-0.1239138	0.169150
##	6292	0.0258531	-0.0323136	-0.1490061	0.143065
##	6302	-0.0369061	0.0934506	0.0161725	0.149652
##	6317	0.0083417	0.1796400	0.0723985	0.060389
##	6322	-0.0369061	0.0934506	0.0161725	0.149652
##	6332	0.0587268	-0.0408445	-0.1786840	0.112212
##	6342	-0.3001578	-0.2187645	-0.0707066	-0.273205
##	6352	-0.4883920	-0.0768245	-0.2602877	-0.334602
##	6362	0.0556963	-0.1848458	0.0163818	0.074160
##	6372	0.1288040	-0.2038175	-0.0496188	0.005546
##	6382	0.0587268	-0.0408445	-0.1786840	0.112212
##	6392	-0.0662518	-0.0084121	-0.0658552	0.229508
##	6402	-0.0662518	-0.0084121	-0.0658552	0.229508
##	6417	-0.1890471	-0.3012971	0.1677070	-0.092952

##	6422	-0.0662518	-0.0084121	-0.0658552	0.229508
##	6432	-0.0036366	0.2089330	0.1091679	0.059119
##	6442	0.0226141	0.3000522	0.1825442	-0.012315
##	6452	-0.1820582	-0.0583812	-0.2201018	-0.093614
##	6462	0.0556963	-0.1848458	0.0163818	0.074160
##	6472	0.0910841	0.3876217	0.2251470	-0.126912
##	6482	0.2958442	-0.4512319	0.0706535	-0.208867
##	6492	0.1708656	-0.4187995	0.1834823	-0.091570
##	6502	0.1708656	-0.4187995	0.1834823	-0.091570
##	6516	-0.0472547	-0.0133419	-0.0830055	0.211679
##	6522	0.1706371	0.1975116	-0.0146562	-0.120596
##	6532	0.1254435	0.5861147	0.3997234	-0.258270
##	6541	-0.0088479	0.1199727	0.0243498	0.107166
##	6552	0.0123891	0.1144616	0.0051773	0.087234
##	6562	0.0258531	-0.0323136	-0.1490061	0.143065
##	6572	0.0244216	0.2354551	0.1173451	0.016632
##	6582	-0.5096290	-0.0713135	-0.2411152	-0.314670
##	6592	0.0123891	0.1144616	0.0051773	0.087234
##	6602	0.3489181	-0.1161503	-0.4406644	-0.160142
##	6616	0.3489181	-0.1161503	-0.4406644	-0.160142
##	6622	0.0456585	0.2299440	0.0981727	-0.003299
##	6632	-0.1834898	0.2093876	0.0462494	-0.220046
##	6642	-0.0260177	-0.0188530	-0.1021779	0.191747
##	6652	0.2239395	-0.0837178	-0.3278356	-0.042846
##	6662	0.3489181	-0.1161503	-0.4406644	-0.160142
##	6672	0.1041524	0.1168332	-0.0517097	-0.011401
##	6682	0.0244216	0.2354551	0.1173451	0.016632
##	6692	0.2239395	-0.0837178	-0.3278356	-0.042846
##	6702	0.3489181	-0.1161503	-0.4406644	-0.160142
##	6716	-0.0472547	-0.0133419	-0.0830055	0.211679
##	6722	0.1508317	-0.0647461	-0.2618349	0.025768
##	6732	0.2239395	-0.0837178	-0.3278356	-0.042846
##	6742	0.2239395	-0.0837178	-0.3278356	-0.042846
##	6752	0.2239395	-0.0837178	-0.3278356	-0.042846
##	6762	0.2239395	-0.0837178	-0.3278356	-0.042846
##	6772	-0.2551660	-0.0394094	-0.1541012	-0.025000
##	6782	0.3489181	-0.1161503	-0.4406644	-0.160142
##	6792	-0.0088479	0.1199727	0.0243498	0.107166
##	6802	0.0537673	0.3373178	0.1993728	-0.063224
##	6816	-0.0472547	-0.0133419	-0.0830055	0.211679
##	6822	-0.0260177	-0.0188530	-0.1021779	0.191747
##	6832	0.0587268	-0.0408445	-0.1786840	0.112212
##	6842	0.0885699	-0.1933766	-0.0132961	0.043307
##	6852	0.2038230	-0.3355913	0.0318378	-0.096584
##	6862	0.2525795	-0.5847923	0.3020420	-0.209158
##	6872	0.0279020	-0.1776330	0.0414741	0.100245
##	6882	0.1673145	-0.1187078	0.2703927	-0.161430
##	6892	0.0313283	0.0903814	-0.0309185	0.078617
##	6902	0.1466225	-0.4900909	0.3084260	-0.090731
##	6916	0.2958442	-0.4512319	0.0706535	-0.208867
##	6922	0.2239395	-0.0837178	-0.3278356	-0.042846
##	6931	0.0712787	0.1253641	-0.0220318	0.019452

## 6941	-0.0019412	-0.0251009	-0.1239138	0.169150
## 6951	0.3249615	0.2669418	-0.0454573	-0.317748
## 6961	-0.0019412	-0.0251009	-0.1239138	0.169150
## 6971	0.0258531	-0.0323136	-0.1490061	0.143065
## 6981	0.3489181	-0.1161503	-0.4406644	-0.160142
## 6991	0.0258531	-0.0323136	-0.1490061	0.143065
## 7001	0.1545573	0.1416965	-0.0596028	-0.076839
## 7015	-0.0019412	-0.0251009	-0.1239138	0.169150
## 7021	-0.0260177	-0.0188530	-0.1021779	0.191747
## 7031	0.3489181	-0.1161503	-0.4406644	-0.160142
## 7041	-0.3300010	-0.0662324	-0.2360945	-0.204300
## 7051	-0.2693330	-0.0819760	-0.2908646	-0.261239
## 7061	-0.3022067	-0.0734451	-0.2611868	-0.230386
## 7071	-0.3300010	-0.0662324	-0.2360945	-0.204300
## 7081	0.1708656	-0.4187995	0.1834823	-0.091570
## 7091	0.0989609	-0.0512854	-0.2150067	0.074451
## 7108	0.1268751	0.3183460	0.1333722	-0.131838
## 7115	0.1268751	0.3183460	0.1333722	-0.131838
## 7121	-0.0019412	-0.0251009	-0.1239138	0.169150
## 7131	0.1652819	0.4516606	0.2407274	-0.236351
## 7141	0.1859994	0.4009344	0.1770706	-0.234125
## 7151	0.3489181	-0.1161503	-0.4406644	-0.160142
## 7161	0.0587268	-0.0408445	-0.1786840	0.112212
## 7171	0.0587268	-0.0408445	-0.1786840	0.112212
## 7181	0.0587268	-0.0408445	-0.1786840	0.112212
## 7191	0.0258531	-0.0323136	-0.1490061	0.143065
## 7201	0.0258531	-0.0323136	-0.1490061	0.143065
## 7215	0.0989609	-0.0512854	-0.2150067	0.074451
## 7221	0.1859994	0.4009344	0.1770706	-0.234125
## 7231	0.0258531	-0.0323136	-0.1490061	0.143065
## 7241	0.0975293	0.2164833	0.0513445	-0.051982
## 7251	0.0258531	-0.0323136	-0.1490061	0.143065
## 7261	0.0587268	-0.0408445	-0.1786840	0.112212
## 7271	0.0258531	-0.0323136	-0.1490061	0.143065
## 7281	0.0885699	-0.1933766	-0.0132961	0.043307
## 7291	-0.3092485	0.3514937	0.0486387	-0.729620
## 7301	0.0587268	-0.0408445	-0.1786840	0.112212
## 7315	0.2239395	-0.0837178	-0.3278356	-0.042846
## 7321	-0.4365212	-0.0902852	-0.3071159	-0.383284
## 7331	0.0258531	-0.0323136	-0.1490061	0.143065
## 7341	0.0258531	-0.0323136	-0.1490061	0.143065
## 7351	0.0587268	-0.0408445	-0.1786840	0.112212
## 7361	0.0975293	0.2164833	0.0513445	-0.051982
## 7371	0.0258531	-0.0323136	-0.1490061	0.143065
## 7381	0.2239395	-0.0837178	-0.3278356	-0.042846
## 7391	0.0258531	-0.0323136	-0.1490061	0.143065
## 7401	0.3489181	-0.1161503	-0.4406644	-0.160142
## 7415	-0.1683110	0.2686837	-0.0084863	-0.536141
## 7421	0.1304030	0.2079525	0.0216666	-0.082835
## 7431	-0.4036475	-0.0988161	-0.3367937	-0.414137
## 7441	0.0587268	-0.0408445	-0.1786840	0.112212
## 7451	0.4264808	0.3632067	0.3202812	-0.581700

## 7461	0.1288040	-0.2038175	-0.0496188	0.005546
## 7471	0.1806748	-0.2172782	-0.0964470	-0.043137
## 7481	0.1288040	-0.2038175	-0.0496188	0.005546
## 7491	0.3533731	0.3821784	0.3862819	-0.513086
## 7501	0.3015023	0.3956391	0.4331100	-0.464403
## 7515	0.1288040	-0.2038175	-0.0496188	0.005546
## 7521	0.1806748	-0.2172782	-0.0964470	-0.043137
## 7531	0.1288040	-0.2038175	-0.0496188	0.005546
## 7541	0.0989609	-0.0512854	-0.2150067	0.074451
## 7551	0.2716591	0.5481712	0.2677221	-0.395498
## 7561	0.0989609	-0.0512854	-0.2150067	0.074451
## 7571	0.1508317	-0.0647461	-0.2618349	0.025768
## 7581	0.1288040	-0.2038175	-0.0496188	0.005546
## 7591	0.1508317	-0.0647461	-0.2618349	0.025768
## 7601	0.2239395	-0.0837178	-0.3278356	-0.042846
## 7615	0.0989609	-0.0512854	-0.2150067	0.074451
## 7621	0.2239395	-0.0837178	-0.3278356	-0.042846
## 7631	0.0989609	-0.0512854	-0.2150067	0.074451
## 7641	0.0989609	-0.0512854	-0.2150067	0.074451
## 7651	0.2716591	0.5481712	0.2677221	-0.395498
## 7661	0.0989609	-0.0512854	-0.2150067	0.074451
## 7671	0.0989609	-0.0512854	-0.2150067	0.074451
## 7681	0.0989609	-0.0512854	-0.2150067	0.074451
## 7691	0.0989609	-0.0512854	-0.2150067	0.074451
## 7701	0.0989609	-0.0512854	-0.2150067	0.074451
## 7715	0.0989609	-0.0512854	-0.2150067	0.074451
## 7721	0.1966948	-0.1190159	0.7998215	-0.372350
## 7731	-0.4729880	-0.0594078	-0.2049393	-0.233505
## 7741	0.1508317	-0.0647461	-0.2618349	0.025768
## 7751	0.2239395	-0.0837178	-0.3278356	-0.042846
## 7761	-0.0260177	-0.0188530	-0.1021779	0.191747
## 7771	0.0989609	-0.0512854	-0.2150067	0.074451
## 7781	0.2239395	-0.0837178	-0.3278356	-0.042846
## 7791	-0.0834368	-0.0039525	-0.0503408	0.245637
## 7801	-0.0662518	-0.0084121	-0.0658552	0.229508
## 7815	-0.0662518	-0.0084121	-0.0658552	0.229508
## 7821	-0.0418151	0.1405214	0.0660008	0.132375
## 7831	-0.1704453	0.0576708	-0.0269093	0.062918
## 7841	0.2239395	-0.0837178	-0.3278356	-0.042846
## 7851	0.3249615	0.2669418	-0.0454573	-0.317748
## 7861	0.1134111	0.4651213	0.2875556	-0.187669
## 7871	-0.0607187	0.1334334	0.0711779	0.155848
## 7881	0.2314250	0.5586121	0.3040449	-0.357737
## 7891	0.1768724	0.4918927	0.2731254	-0.267891
## 7901	-0.0061614	0.2642798	0.1656606	0.035354
## 7915	-0.2385362	0.0362964	0.0896807	0.391203
## 7921	-0.0260177	-0.0188530	-0.1021779	0.191747
## 7931	0.3249615	0.2669418	-0.0454573	-0.317748
## 7941	-0.3779523	-0.0029500	-0.0283882	0.115039
## 7951	0.4976597	0.8663985	0.4372716	-0.787697
## 7961	0.0869247	0.2833634	0.1244855	-0.072673
## 7971	-0.0834368	-0.0039525	-0.0503408	0.245637

## 7981	-0.1079949	0.2420697	0.2093834	0.154169
## 7991	0.0099184	0.3200948	0.2106072	-0.008403
## 8001	0.2225079	0.1840509	-0.0614843	-0.169278
## 8015	-0.0834368	-0.0039525	-0.0503408	0.245637
## 8021	-0.0019412	-0.0251009	-0.1239138	0.169150
## 8031	0.2225079	0.1840509	-0.0614843	-0.169278
## 8041	0.2239395	-0.0837178	-0.3278356	-0.042846
## 8051	0.3966377	0.5157388	0.1548933	-0.512795
## 8061	-0.1343803	0.1868512	0.1716810	0.208590
## 8071	0.1758286	0.3656302	0.1486409	-0.206448
## 8081	-0.1445213	0.1152821	0.1072843	0.253565
## 8091	0.0033281	0.0830097	-0.0201502	0.111891
## 8108	-0.2032286	0.1324717	0.1622220	0.307729
## 8115	0.2225079	0.1840509	-0.0614843	-0.169278
## 8121	0.3249615	0.2669418	-0.0454573	-0.317748
## 8131	-0.0976889	0.2393952	0.2000793	0.144496
## 8141	-0.1391549	0.2987921	0.2857251	0.160173
## 8151	-0.0061614	0.2642798	0.1656606	0.035354
## 8161	0.2239395	-0.0837178	-0.3278356	-0.042846
## 8171	-2.0171076	-0.0771363	-0.5224288	-1.980519
## 8181	-0.0662518	-0.0084121	-0.0658552	0.229508
## 8191	0.4976597	0.8663985	0.4372716	-0.787697
## 8201	-0.0233511	0.2046124	0.1176118	0.082130
## 8214	-0.0019412	-0.0251009	-0.1239138	0.169150
## 8221	-0.0260177	-0.0188530	-0.1021779	0.191747
## 8231	0.2239395	-0.0837178	-0.3278356	-0.042846
## 8241	-0.0834368	-0.0039525	-0.0503408	0.245637
## 8251	-0.0834368	-0.0039525	-0.0503408	0.245637
## 8261	0.2239395	-0.0837178	-0.3278356	-0.042846
## 8271	-0.1722333	0.0190905	0.0298233	0.328975
## 8281	-0.0019412	-0.0251009	-0.1239138	0.169150
## 8291	-0.7664704	0.0090828	-0.0993325	-0.351817
## 8301	0.3249615	0.2669418	-0.0454573	-0.317748
## 8315	0.0949470	0.4010303	0.2359446	-0.137424
## 8321	-0.3338424	-0.0492737	-0.1810047	-0.114562
## 8331	-0.1097301	0.0669988	0.0369637	0.239670
## 8341	-0.1790743	0.0662157	0.0809526	0.313725
## 8351	0.0279020	-0.1776330	0.0414741	0.100245
## 8361	0.4837012	-0.8050441	0.3062911	-0.471344
## 8371	-0.1160138	-0.4116386	0.7116271	0.099745
## 8381	-0.0260177	-0.0188530	-0.1021779	0.191747
## 8391	-0.2348715	0.2510474	0.1842448	-0.018970
## 8401	-0.0950936	0.0679550	0.0284629	0.223661
## 8415	-0.0514639	-0.4386867	0.4872555	0.095179
## 8421	-0.0767444	0.2279047	0.1751685	0.127733
## 8431	0.0033281	0.0830097	-0.0201502	0.111891
## 8441	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8451	0.2239395	-0.0837178	-0.3278356	-0.042846
## 8461	-0.1522151	-0.2109133	-0.0547139	-0.162519
## 8471	-0.0662518	-0.0084121	-0.0658552	0.229508
## 8481	-0.1393596	0.0105597	0.0001455	0.298122
## 8491	-0.1393596	0.0105597	0.0001455	0.298122

## 8501	-0.2084154	0.0284799	0.0624880	0.362933
## 8514	-0.1619273	0.0164161	0.0205192	0.319303
## 8521	0.1508317	-0.0647461	-0.2618349	0.025768
## 8531	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8541	0.2239395	-0.0837178	-0.3278356	-0.042846
## 8551	-0.0834368	-0.0039525	-0.0503408	0.245637
## 8561	0.1041524	0.1168332	-0.0517097	-0.011401
## 8571	0.1041524	0.1168332	-0.0517097	-0.011401
## 8581	0.0258531	-0.0323136	-0.1490061	0.143065
## 8591	-0.0117606	0.2448445	0.1500098	0.050590
## 8601	-0.0019412	-0.0251009	-0.1239138	0.169150
## 8615	0.0175805	0.2825803	0.1684744	0.001382
## 8621	-0.1258621	0.2509217	0.2296921	0.168923
## 8631	-0.4274771	0.0201854	0.0495800	0.217012
## 8641	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8651	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8661	0.3249615	0.2669418	-0.0454573	-0.317748
## 8671	0.0438312	0.3736995	0.2418507	-0.070052
## 8681	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8691	-0.0472547	-0.0133419	-0.0830055	0.211679
## 8701	0.0018965	0.3507784	0.2462010	-0.014542
## 8714	-0.0208262	0.1492657	0.0611191	0.105895
## 8721	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8731	0.0244216	0.2354551	0.1173451	0.016632
## 8741	0.2383897	0.4326889	0.1747268	-0.304965
## 8751	0.1597488	0.3098152	0.1036943	-0.162691
## 8761	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8771	0.0175805	0.2825803	0.1684744	0.001382
## 8781	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8791	-0.0991255	0.0001188	-0.0361773	0.260361
## 8801	-0.0019412	-0.0251009	-0.1239138	0.169150
## 8815	0.0587268	-0.0408445	-0.1786840	0.112212
## 8821	-0.1393596	0.0105597	0.0001455	0.298122
## 8831	0.1531258	0.4094653	0.2067485	-0.203272
## 8841	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8851	-0.1483297	0.2370449	0.2304406	0.199427
## 8861	0.3249615	0.2669418	-0.0454573	-0.317748
## 8871	-0.0834368	-0.0039525	-0.0503408	0.245637
## 8881	0.0869247	0.2833634	0.1244855	-0.072673
## 8891	0.1041524	0.1168332	-0.0517097	-0.011401
## 8901	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8915	0.5216163	0.4833064	0.0420645	-0.630091
## 8921	0.0936666	0.5466844	0.3811515	-0.205664
## 8931	-0.0662518	-0.0084121	-0.0658552	0.229508
## 8941	0.1999829	0.2993742	0.0673716	-0.200452
## 8951	-0.0267767	0.3050912	0.2194233	0.037757
## 8961	0.2225079	0.1840509	-0.0614843	-0.169278
## 8971	0.3489181	-0.1161503	-0.4406644	-0.160142
## 8981	0.6226383	0.8339660	0.3244428	-0.904994
## 8991	-0.1025679	0.2675163	0.1273894	-0.135390
## 9001	0.3489181	-0.1161503	-0.4406644	-0.160142
## 9015	-0.1157978	0.4078404	0.3190640	-0.073052

##	9021	-0.0662518	-0.0084121	-0.0658552	0.229508
##	9031	-0.2640535	0.0215041	0.0434625	0.299597
##	9041	0.1508317	-0.0647461	-0.2618349	0.025768
##	9051	-0.0019412	-0.0251009	-0.1239138	0.169150
##	9061	-0.0418151	0.1405214	0.0660008	0.132375
##	9071	-0.5782799	-0.0399925	-0.1354595	-0.177360
##	9081	-0.0662518	-0.0084121	-0.0658552	0.229508
##	9091	-0.0019412	-0.0251009	-0.1239138	0.169150
##	9107	-0.0260177	-0.0188530	-0.1021779	0.191747
##	9115	0.3489181	-0.1161503	-0.4406644	-0.160142
##	9121	0.3489181	-0.1161503	-0.4406644	-0.160142
##	9131	-0.3060277	0.0532963	0.0562797	0.209070
##	9141	-0.1712806	0.2951353	0.3028390	0.196055
##	9151	0.3249615	0.2669418	-0.0454573	-0.317748
##	9161	-0.1343803	0.1868512	0.1716810	0.208590
##	9171	-0.0834368	-0.0039525	-0.0503408	0.245637
##	9181	-0.0779040	0.1276223	0.0765117	0.176885
##	9191	0.5216163	0.4833064	0.0420645	-0.630091
##	9201	0.3489181	-0.1161503	-0.4406644	-0.160142
##	9214	-0.0767444	0.2279047	0.1751685	0.127733
##	9221	0.3489181	-0.1161503	-0.4406644	-0.160142
##	9231	-0.2401065	0.0606502	0.0616125	0.241841
##	9241	-0.1722333	0.0190905	0.0298233	0.328975
##	9251	-0.0472547	-0.0133419	-0.0830055	0.211679
##	9261	-0.1476862	0.0658486	0.0603261	0.280550
##	9271	0.1508317	-0.0647461	-0.2618349	0.025768
##	9281	0.2239395	-0.0837178	-0.3278356	-0.042846
##	9291	-0.4729880	-0.0594078	-0.2049393	-0.233505
##	9301	-0.0418151	0.1405214	0.0660008	0.132375
##	9314	-0.0208262	0.1492657	0.0611191	0.105895
##	9321	-0.0019412	-0.0251009	-0.1239138	0.169150
##	9331	0.2239395	-0.0837178	-0.3278356	-0.042846
##	9341	0.0587268	-0.0408445	-0.1786840	0.112212
##	9351	0.0989609	-0.0512854	-0.2150067	0.074451
##	9361	-0.5979666	-0.0269754	-0.0921105	-0.116208
##	9371	-0.0260177	-0.0188530	-0.1021779	0.191747
##	9381	0.3489181	-0.1161503	-0.4406644	-0.160142
##	9391	-0.0472547	-0.0133419	-0.0830055	0.211679
##	9401	-0.0550151	-0.3601826	0.3874040	0.120426
##	9413	0.4738140	-0.7545096	0.2514974	-0.448515
##	9421	0.0258531	-0.0323136	-0.1490061	0.143065
##	9431	0.3139491	-0.2831218	0.6989618	-0.463589
##	9441	-0.0991255	0.0001188	-0.0361773	0.260361
##	9451	0.1545573	0.1416965	-0.0596028	-0.076839
##	9461	0.1806748	-0.2172782	-0.0964470	-0.043137
##	9471	-0.1619273	0.0164161	0.0205192	0.319303
##	9481	-0.0210040	0.0777774	-0.0096292	0.140245
##	9491	-0.0662518	-0.0084121	-0.0658552	0.229508
##	9501	-0.1619273	0.0164161	0.0205192	0.319303
##	9513	-0.0260177	-0.0188530	-0.1021779	0.191747
##	9521	-0.0260177	-0.0188530	-0.1021779	0.191747
##	9531	0.0258531	-0.0323136	-0.1490061	0.143065

## 9541	-0.1135577	0.0038640	-0.0231481	0.273906
## 9551	0.1508317	-0.0647461	-0.2618349	0.025768
## 9561	0.3489181	-0.1161503	-0.4406644	-0.160142
## 9571	-0.0950936	0.0679550	0.0284629	0.223661
## 9581	-0.0617578	0.0712978	0.0102566	0.186644
## 9591	-0.0991255	0.0001188	-0.0361773	0.260361
## 9601	-0.1819819	0.0216203	0.0386243	0.338124
## 9613	-0.0472547	-0.0133419	-0.0830055	0.211679
## 9621	0.0313283	0.0903814	-0.0309185	0.078617
## 9631	0.2239395	-0.0837178	-0.3278356	-0.042846
## 9641	-0.1667581	0.1417856	0.1479109	0.264528
## 9651	-0.0662518	-0.0084121	-0.0658552	0.229508
## 9661	-0.1820582	-0.0583812	-0.2201018	-0.093614
## 9671	-0.1135577	0.0038640	-0.0231481	0.273906
## 9681	-0.1358916	0.0660096	0.0528717	0.267940
## 9691	-0.2000276	0.0263033	0.0549156	0.355061
## 9701	-0.2164303	0.0305598	0.0697238	0.370455
## 9714	-0.0472547	-0.0133419	-0.0830055	0.211679
## 9721	0.1508317	-0.0647461	-0.2618349	0.025768
## 9731	-0.0472547	-0.0133419	-0.0830055	0.211679
## 9741	0.0989609	-0.0512854	-0.2150067	0.074451
## 9751	-0.0472547	-0.0133419	-0.0830055	0.211679
## 9761	-0.2223962	0.1348098	0.1288854	0.201191
## 9771	0.3489181	-0.1161503	-0.4406644	-0.160142
## 9781	-0.0472547	-0.0133419	-0.0830055	0.211679
## 9791	-0.0472547	-0.0133419	-0.0830055	0.211679
## 9801	0.0027426	0.1602048	0.0567477	0.075625
## 9814	-0.2650320	0.2254978	0.2005105	0.083224
## 9821	0.0587268	-0.0408445	-0.1786840	0.112212
## 9831	0.0443643	0.3046787	0.1730893	-0.037636
## 9841	0.0099184	0.3200948	0.2106072	-0.008403
## 9851	-0.2122896	0.1010197	0.1368944	0.332386
## 9861	0.0697351	0.2236961	0.0764367	-0.025896
## 9871	-0.0866137	0.2778386	0.2310369	0.114358
## 9881	-0.1269198	0.0073315	-0.0110850	0.286447
## 9891	-0.1194169	0.0797833	0.0558897	0.243854
## 9901	-0.8561968	-0.0793159	-0.2449783	-0.517897
## 9914	-0.1135577	0.0038640	-0.0231481	0.273906
## 9921	-0.0425052	0.0740153	0.0005218	0.164888
## 9931	0.0587268	-0.0408445	-0.1786840	0.112212
## 9941	0.0989609	-0.0512854	-0.2150067	0.074451
## 9951	0.1508317	-0.0647461	-0.2618349	0.025768
## 9961	-0.0260177	-0.0188530	-0.1021779	0.191747
## 9971	0.0258531	-0.0323136	-0.1490061	0.143065
## 9981	-0.0766268	0.3873210	0.3331147	0.051430
## 9991	-0.8566859	-0.1075151	-0.3361455	-0.670291
## 10001	0.0800180	0.4284370	0.2727491	-0.134658
## 10014	0.0508500	0.3980626	0.2614697	-0.089151
## 10021	-0.0662518	-0.0084121	-0.0658552	0.229508
## 10031	0.0989609	-0.0512854	-0.2150067	0.074451
## 10041	-0.2447904	0.1542571	0.2106471	0.341480
## 10051	-0.0112297	-0.4491276	0.4509327	0.057418

##	10061	-0.0950936	0.0679550	0.0284629	0.223661
##	10071	-0.5871107	-0.3759575	-0.0484614	-0.679659
##	10081	0.1856535	-0.8490742	0.7365930	-0.225899
##	10091	0.2525795	-0.5847923	0.3020420	-0.209158
##	10108	0.0038255	-0.1713851	0.0632100	0.122842
##	10114	-0.1865872	-0.1219723	0.2351117	0.301550
##	10121	0.0989609	-0.0512854	-0.2150067	0.074451
##	10131	0.0699635	-0.3926150	0.2745752	0.003129
##	10141	0.3968661	-0.1005723	0.3530318	-0.483769
##	10151	-0.0008835	0.2184892	0.1168633	0.051627
##	10161	0.3489181	-0.1161503	-0.4406644	-0.160142
##	10171	-0.0278450	0.1249025	0.0415001	0.124995
##	10181	-0.3758086	0.0105773	0.0152241	0.189025
##	10191	-0.3207147	-0.0403161	-0.1528692	-0.060163
##	10201	-0.0920998	0.2180163	0.1752793	0.148774
##	10214	0.0989609	-0.0512854	-0.2150067	0.074451
##	10221	0.0163313	-0.0227102	0.6758675	-0.116644
##	10231	-0.0472547	-0.0133419	-0.0830055	0.211679
##	10241	0.0258531	-0.0323136	-0.1490061	0.143065
##	10251	-0.0662518	-0.0084121	-0.0658552	0.229508
##	10261	-0.0674549	-0.3569544	0.3986345	0.132101
##	10271	0.0989609	-0.0512854	-0.2150067	0.074451
##	10281	-0.1973402	0.0669233	0.0934456	0.332795
##	10291	-0.0662518	-0.0084121	-0.0658552	0.229508
##	10301	-0.1912304	0.0240204	0.0469737	0.346804
##	10314	0.0533253	0.2565564	0.1196030	-0.024162
##	10321	0.2239395	-0.0837178	-0.3278356	-0.042846
##	10331	-0.1506671	-0.2023499	0.6056047	0.146639
##	10341	-0.0936503	0.1228139	0.0819103	0.195914
##	10351	0.2225079	0.1840509	-0.0614843	-0.169278
##	10361	-0.0774465	0.0753691	0.0244201	0.201368
##	10371	0.1508317	-0.0647461	-0.2618349	0.025768
##	10381	0.0989609	-0.0512854	-0.2150067	0.074451
##	10391	-0.1777277	-0.0471407	-0.1840244	-0.030958
##	10401	0.2239395	-0.0837178	-0.3278356	-0.042846
##	10414	0.0258531	-0.0323136	-0.1490061	0.143065
##	10421	-0.3917086	0.0222348	0.0539354	0.244589
##	10431	0.0611714	-0.0621507	0.1344694	0.009712
##	10441	0.0989609	-0.0512854	-0.2150067	0.074451
##	10451	0.0018965	0.3507784	0.2462010	-0.014542
##	10461	-0.0435291	0.1931007	0.1192267	0.109071
##	10471	0.0584624	-0.6546538	0.9249049	-0.155300
##	10481	0.0338508	-0.0517183	0.6358034	-0.121397
##	10491	-0.0008835	0.2184892	0.1168633	0.051627
##	10501	-0.1391549	0.2987921	0.2857251	0.160173
##	10514	-0.1393596	0.0105597	0.0001455	0.298122
##	10521	-1.0084701	-0.0984077	-0.2970485	-0.691239
##	10531	-0.0208262	0.1492657	0.0611191	0.105895
##	10541	0.0033281	0.0830097	-0.0201502	0.111891
##	10551	-0.1944855	0.1628198	0.1866606	0.283938
##	10561	-0.4622435	0.1271368	-0.0070750	-0.281976
##	10571	-0.5619888	-0.0860520	-0.2854542	-0.418382

```

## 10581 0.0989609 -0.0512854 -0.2150067 0.074451
## 10591 -0.6869674 -0.0536195 -0.1726254 -0.301085
## 10601 -0.0260177 -0.0188530 -0.1021779 0.191747
## 10614 -0.0879724 -0.2218032 0.7258104 0.030333
## 10621 -0.1119785 -0.5759743 0.7469526 -0.067700
## 10631 -0.0828587 -0.5137840 0.7264762 0.068256
## 10641 -0.0425052 0.0740153 0.0005218 0.164888
## 10651 -0.1547040 0.1508079 0.1490728 0.247408
## 10661 -0.0472547 -0.0133419 -0.0830055 0.211679
## 10671 -0.0260177 -0.0188530 -0.1021779 0.191747
## 10681 -0.0260177 -0.0188530 -0.1021779 0.191747
## 10691 -0.1630550 0.1646484 0.0952330 0.058608
## 10701 0.0295787 0.1741290 0.0532261 0.040457
## 10714 0.0123891 0.1144616 0.0051773 0.087234
## 10721 0.0587268 -0.0408445 -0.1786840 0.112212
## 10731 -0.0950936 0.0679550 0.0284629 0.223661
## 10741 -0.0618241 0.1834374 0.1214583 0.133128
## 10751 -0.1762809 0.1087307 0.1212925 0.290441
## 10761 -0.3122727 -0.0326435 -0.1285917 -0.014861
## 10771 0.0533253 0.2565564 0.1196030 -0.024162
## 10781 -0.3230979 0.0888659 0.1260366 0.271702
## 10791 -0.1393596 0.0105597 0.0001455 0.298122
## 10801 -0.1135577 0.0038640 -0.0231481 0.273906
## 10814 -0.2702504 0.0445264 0.1183117 0.420967
## 10821 0.0587268 -0.0408445 -0.1786840 0.112212
## 10831 -0.6583445 -0.0458237 -0.1492316 -0.245799
## 10841 0.0258531 -0.0323136 -0.1490061 0.143065
## 10851 -0.2084154 0.0284799 0.0624880 0.362933
## 10861 -0.0472547 -0.0133419 -0.0830055 0.211679
## 10871 -0.1984882 0.1119604 0.0709148 0.134622
## 10881 0.0258531 -0.0323136 -0.1490061 0.143065
## 10891 -0.1864002 -0.0234760 -0.1069401 0.092736

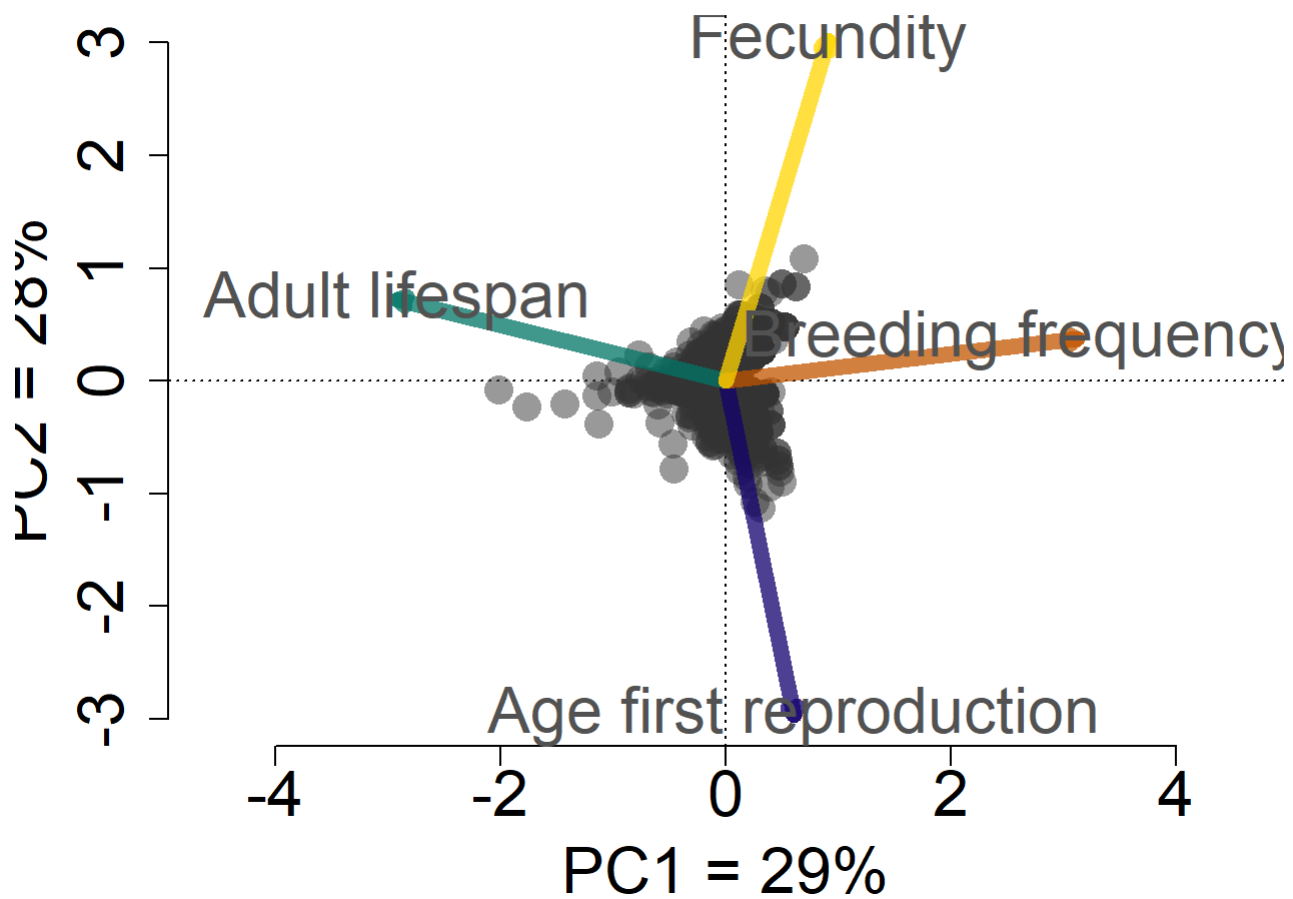
```

Figure

```

par(mar=c(4,4,2,2), bty="n")
plot(test.pca_oystercatcher1, scaling = scl, choices=c(1,2), type="n", main="", ylim=c(-3,3), x
lim=c(-3.5,3.5), xlab="PC1 = 29%", ylab="PC2 = 28%", cex.lab=2, cex.axis=2)
with(data_oystercatcher, points(test.pca_oystercatcher1, display = "sites", col = col_points, sc
aling = scl, pch = 16, cex=2))
g <- vegan::scores(test.pca_oystercatcher1, display = "species")
len <- 1
arrows(0, 0, len * g[, 1], len * g[, 2], length = 0.05, col = c(col_afr, col_life, col_breed, c
ol_fec), lwd=8)
text(x=g[1,1], y=g[1,2], labels="Age first reproduction", col = "gray32", cex=2)
text(x=g[2,1], y=g[2,2], labels="Adult lifespan", col = "gray32", cex=2)
text(x=g[3,1]-0.5, y=g[3,2], labels="Breeding frequency", col = "gray32", cex=2)
text(x=g[4,1], y=g[4,2], labels="Fecundity", col = "gray32", cex=2)

```



```
# Test for proportion of variance explained per axe
sum_all_oystercatcher <- summary(test.pca_oystercatcher1)

prop_oystercatcher <- data.frame(
  PC = c(1:4),
  prim = c(1:4),
  adult = c(1:4),
  breed = c(1:4),
  freq = c(1:4)
)

for (i in 1:4) {
  for (j in 1:4){
    prop_oystercatcher[j,i+1] <- as.numeric(abs(sum_all_oystercatcher$species[i,j])/sum(abs(sum_
all_oystercatcher$species[i,])))
  }
}

prop_oystercatcher
```

```
##      PC      prim      adult      breed      freq
## 1  1 0.08580355 0.38839510 0.46250494 0.1204486
## 2  2 0.41987399 0.09649716 0.05551811 0.4002845
## 3  3 0.35773304 0.26378659 0.11898052 0.2532864
## 4  4 0.13658942 0.25132115 0.36299643 0.2259804
```

Covariances

```
test <- as.matrix(scale(log(data_oystercatcher[,c(2,3,4,7)])))
cov(test)
```

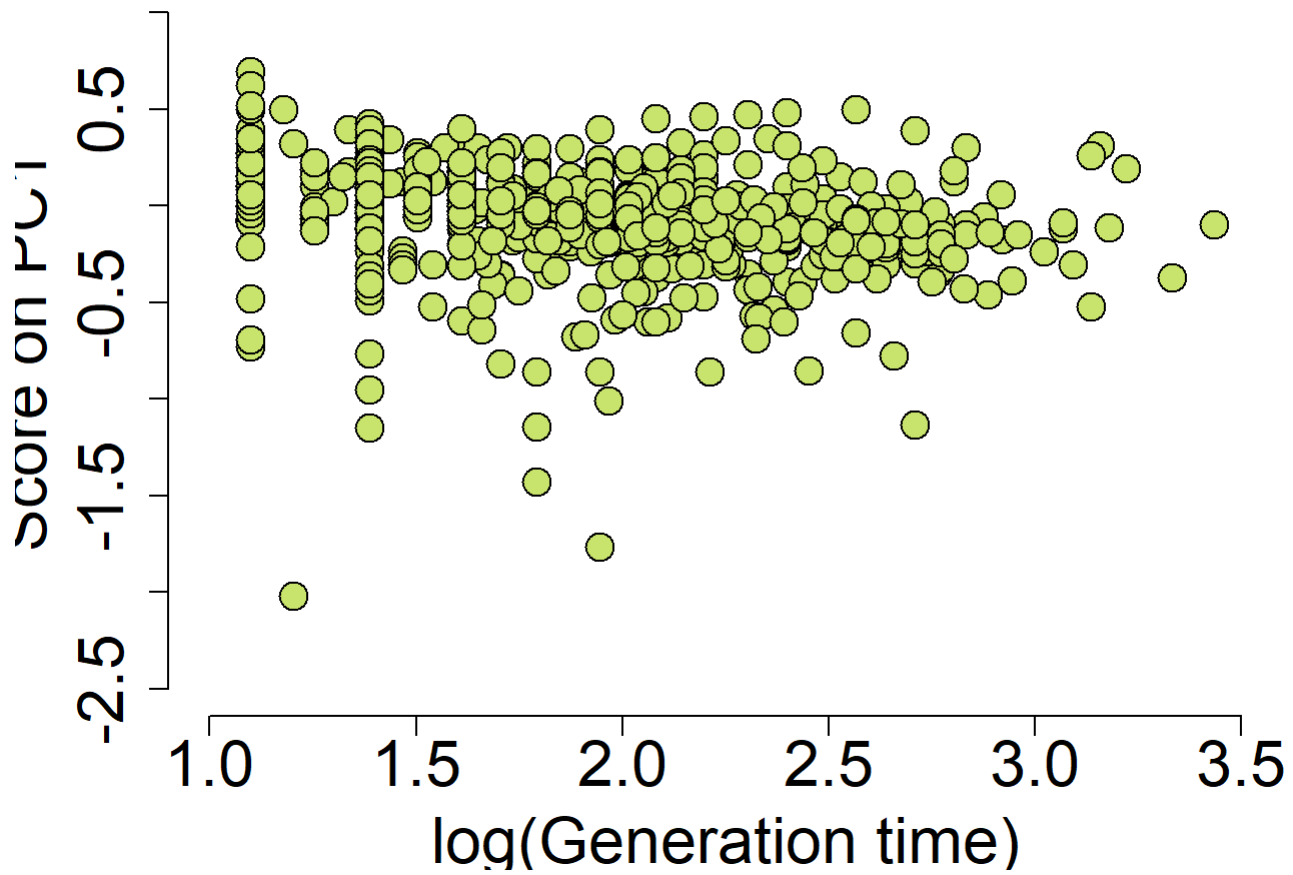
```
##           age_first_repro lifetime_adult breed_freq fec_one
## age_first_repro      1.00000000 -0.044383768 0.02702281 -0.117165603
## lifetime_adult      -0.04438377  1.000000000 -0.15845733 0.006397973
## breed_freq          0.02702281 -0.158457329 1.00000000 0.080264716
## fec_one             -0.11716560  0.006397973 0.08026472 1.000000000
```

Correlation with generation time

```
individual_score_oystercatcher <- as.numeric(vegan::scores(test.pca_oystercatcher1)$sites[,1])
summary(lm(individual_score_oystercatcher ~ log(data_oystercatcher$generation_time)))
```

```
##
## Call:
## lm(formula = individual_score_oystercatcher ~ log(data_oystercatcher$generation_time))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.14874 -0.05540  0.01458  0.10284  0.66305
##
## Coefficients:
##                                Estimate Std. Error t value Pr(>|t|)
## (Intercept)                   0.39140     0.02583   15.15  <2e-16
## log(data_oystercatcher$generation_time) -0.21576     0.01375  -15.70  <2e-16
##
## (Intercept)                  ***
## log(data_oystercatcher$generation_time) ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2224 on 1087 degrees of freedom
## Multiple R-squared:  0.1848, Adjusted R-squared:  0.184
## F-statistic: 246.4 on 1 and 1087 DF, p-value: < 2.2e-16
```

```
par(bty="n")
plot(log(data_oystercatcher$generation_time), individual_score_oystercatcher, pch=21, xlab="log
(Generation time)", ylab="Score on PC1", xlim=c(1,3.5), ylim=c(-2.5,1), bg=colors[11], cex=2, ce
x.axis=2, cex.lab=2)
```



Correlation between species generation time and

percentage of variance on PC1

```
# Compile all proportion of variance on PC1
# Proportion of variance on PC1
ev_squirrels <- test.pca_squirrels1$CA$eig
pc1_squirrels <- as.numeric(ev_squirrels[1])/sum(as.numeric(ev_squirrels))

ev_dipper <- test.pca_dipper1$CA$eig
pc1_dipper <- as.numeric(ev_dipper[1])/sum(as.numeric(ev_dipper))

ev_goats <- test.pca_goats1$CA$eig
pc1_goats <- as.numeric(ev_goats[1])/sum(as.numeric(ev_goats))

ev_sheep <- test.pca_sheep1$CA$eig
pc1_sheep <- as.numeric(ev_sheep[1])/sum(as.numeric(ev_sheep))

ev_fulmar <- test.pca_fulmar1$CA$eig
pc1_fulmar <- as.numeric(ev_fulmar[1])/sum(as.numeric(ev_fulmar))

ev_oystercatcher <- test.pca_oystercatcher1$CA$eig
pc1_oystercatcher <- as.numeric(ev_oystercatcher[1])/sum(as.numeric(ev_oystercatcher))

ev_petrel <- test.pca_petrel1$CA$eig
pc1_petrel <- as.numeric(ev_petrel[1])/sum(as.numeric(ev_petrel))

ev_WA <- test.pca_WA1$CA$eig
pc1_WA <- as.numeric(ev_WA[1])/sum(as.numeric(ev_WA))

ev_bbalbatross <- test.pca_bbalbatross1$CA$eig
pc1_bbalbatross <- as.numeric(ev_bbalbatross[1])/sum(as.numeric(ev_bbalbatross))

ev_btits <- test.pca_btits1$CA$eig
pc1_btits <- as.numeric(ev_btits[1])/sum(as.numeric(ev_btits))

ev_gtits <- test.pca_gtits1$CA$eig
pc1_gtits <- as.numeric(ev_gtits[1])/sum(as.numeric(ev_gtits))

ev_sparrow <- test.pca_sparrow1$CA$eig
pc1_sparrow <- as.numeric(ev_sparrow[1])/sum(as.numeric(ev_sparrow))

ev_deer <- test.pca_deer1$CA$eig
pc1_deer <- as.numeric(ev_deer[1])/sum(as.numeric(ev_deer))

ev_seal <- test.pca_seal1$CA$eig
pc1_seal <- as.numeric(ev_seal[1])/sum(as.numeric(ev_seal))

ev_humans <- test.pca_humans1$CA$eig
pc1_humans <- as.numeric(ev_humans[1])/sum(as.numeric(ev_humans))

ev_gull <- test.pca_gull1$CA$eig
pc1_gull <- as.numeric(ev_gull[1])/sum(as.numeric(ev_gull))
```

```

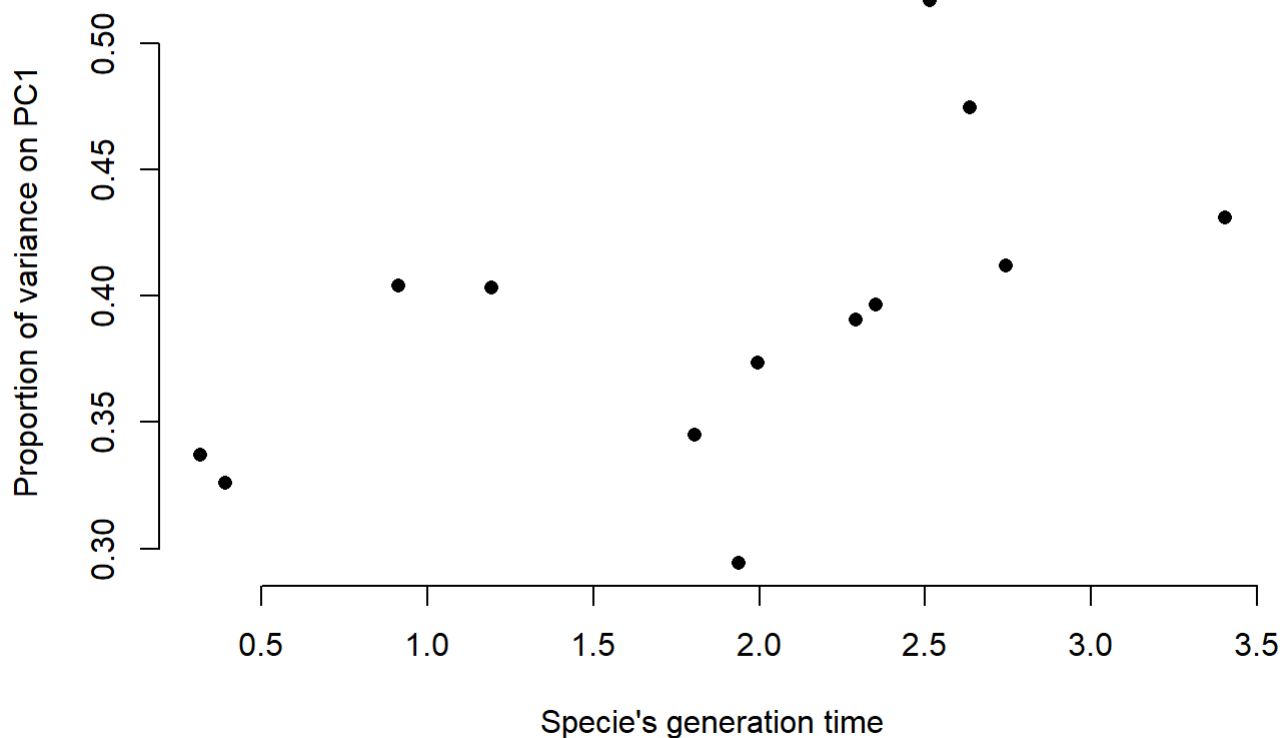
ev_marmots <- test.pca_marmots1$CA$eig
pc1_marmots <- as.numeric(ev_marmots[1])/sum(as.numeric(ev_marmots))

pc1 <- c(pc1_fulmar, pc1_sheep, pc1_bbalbatross, pc1_btits, pc1_dipper, pc1_squirrels, pc1_gtits,
pc1_sparrow, pc1_humans, pc1_goats, pc1_oystercatcher, pc1_gull, pc1_deer, pc1_petrel, pc1_WA, pc1_seal, pc1_marmots)

# Remove great tits, blue tits, Weddell seal, golden-mantled ground squirrel
pc1 <- pc1[-c(4,6,7,16)]

par(bty="n")
plot(log(table_summary[-c(4,6,7,16), "mean_gentime"]), pc1, pch=16, xlab="Specie's generation time",
ylab="Proportion of variance on PC1")

```



```

test <- lm(pc1~log(table_summary[-c(4,6,7,16), "mean_gentime"]))
summary(test)

```

```
##
## Call:
## lm(formula = pc1 ~ log(table_summary[-c(4, 6, 7, 16), "mean_gentime"]))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.10055 -0.01695 -0.01280  0.03575  0.10140
##
## Coefficients:
##                                Estimate Std. Error t value
## (Intercept)                   0.32458    0.03386   9.585
## log(table_summary[-c(4, 6, 7, 16), "mean_gentime"]) 0.03618    0.01622   2.230
##                                Pr(>|t|)
## (Intercept)                   1.13e-06 ***
## log(table_summary[-c(4, 6, 7, 16), "mean_gentime"]) 0.0475 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.0526 on 11 degrees of freedom
## Multiple R-squared:  0.3114, Adjusted R-squared:  0.2488
## F-statistic: 4.974 on 1 and 11 DF, p-value: 0.04751
```

Comparison of proportion of variance explained by

PC among and within species

```
# need to remove great tits, blue tits, squirrels and weddell seals
pc_imp <- data.frame(
  species = c(1:13),
  PC1 = c(1:13),
  PC2 = c(1:13),
  PC3 = c(1:13),
  PC4 = c(1:13)
)

for(i in 1:4){
pc_imp[,i+1] <- c(sum_all_fulmar$cont$importance[2,i],
  sum_all_sheep$cont$importance[2,i],
  sum_all_bbalbatross$cont$importance[2,i],
  #sum_all_btits$cont$importance[2,i],
  sum_all_dipper$cont$importance[2,i],
  #sum_all_squirrels$cont$importance[2,i],
  sum_all_sparrow$cont$importance[2,i],
  sum_all_humans$cont$importance[2,i],
  sum_all_goats$cont$importance[2,i],
  sum_all_oystercatcher$cont$importance[2,i],
  sum_all_gull$cont$importance[2,i],
  sum_all_deer$cont$importance[2,i],
  sum_all_petrel$cont$importance[2,i],
  sum_all_WA$cont$importance[2,i],
  #sum_all_seal$cont$importance[2,i],
  sum_all_marmots$cont$importance[2,i])
}

# interspecific level - using pPCA with average tree
sum_all_pPCA <- c(0.77, 0.17, 0.03, 0.03)

par(mar=c(3,5,2,2))
barplot(sum_all_pPCA, beside=T, ylim=c(0,1), border=NA, width = 0.5, space = 2, col=col_life, ylab="Relative proportion \n of variation explained", names.arg=c("", "", "", ""))
barplot(as.matrix(pc_imp[,c(2:5)])/nrow(pc_imp), ylim=c(0,1), col=colors[c(1:3,5,8:15,17)], border=NA, width = 0.5, space = c(3,2,2,2), add=T, axes=F, names.arg=c("", "", "", ""))
axis(1, at=c(1.5,3,4.5,6), labels=c("PC1", "PC2", "PC3", "PC4"))
```

